Divide-and-conquer: Approaching the capacity of the two-pair bidirectional Gaussian relay network

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Abstract—The capacity region of multi-pair bidirectional relay networks, in which a relay node facilitates the communication between multiple pairs of users, is studied. This problem is first examined in the context of the linear shift deterministic channel model. The capacity region of this network when the relay is operating at either full-duplex mode or half-duplex mode for arbitrary number of pairs is characterized. It is shown that the cut-set upper-bound is tight and the capacity region is achieved by a so called divide-and-conquer relaying strategy. The insights gained from the deterministic network are then used for the Gaussian bidirectional relay network. The strategy in the deterministic channel translates to a specific superposition of lattice codes and random Gaussian codes at the source nodes and successive interference cancelation at the receiving nodes for the Gaussian network. The achievable rate of this scheme with two pairs is analyzed and it is shown that for all channel gains it achieves to within 3 bits/sec/Hz per user of the cut-set upper-bound. Hence, the capacity region of the two-pair bidirectional Gaussian relay network to within 3 bits/sec/Hz per user is characterized.

Index Terms—Bidirectional communication, capacity region, deterministic approach, multi-pair relay network, two-way

I. INTRODUCTION

Cooperative communication and relaying is one of the important research topics in wireless network information theory. The basic model to study this problem is the 3-node relay channel which was first introduced in 1971 by van der Meulen [4] and several strategies for this network were developed by Cover and El Gamal [5].

While the main focus so far has been on the one-way-relay channel, bidirectional communication has also attracted attention. Bidirectional (or two-way) communication between two nodes was first studied by Shannon himself in [6]. Nowadays the bidirectional communication where an additional node acting as a relay is supporting the exchange of information between the two nodes (or one pair) is gaining increased attention. Some relaying strategies for the one-pair bidirectional relay channel, such as decode-and-forward, compress-and-forward and amplify-and-forward, have been analyzed in [7]. An interesting strategy referred to as noisy network coding was proposed in [8], which generalizes the compress-and-forward strategy in [5].

Network coding type techniques have been proposed also by [9], [10], [11], [12] (and others) in order to improve the transmission rate. In [9], a network coding approach is used for the first time in a wireless network in order to reduce the number of transmissions needed to exchange the number of data packets between two nodes of bidirectional setup. While before 4 transmissions were needed, the number of transmissions was to reduce to 3 in [9] resulting in higher data rates. The transmit strategy in [10] is similar to [9] with the extension that a channel code is used by the nodes when communicating to the relay. Once the data is received at the nodes, they perform iterative network and channel decoding resulting in higher rates than without network coding. In [11], [12] the number of transmissions is further reduced by allowing the nodes to submit their data simultaneously to the relay resulting in a multiple-access setup. Additionally, [11], [12] utilize the idea of network coding for the binary case to extend it to the Gaussian case by using lattice coding, which is referred to as physical layer network coding. In [12], it is shown that the lattice based scheme outperforms other schemes at high SNR. It turns out, however, that decoding the individual data streams in the multiple-access hop gives better performance at lower SNR. In [11] decode-and-forward, amplify-and-forward, and modulo-and-forward relaying strategies are compared in terms of transmission rate. It turns out that depending on the scenario, one of schemes outperforms the other two, i.e., neither one is always outperforming the other. The tightest gap characterization on the capacity for the two-way relay channel is provided in [13], where it is shown that upper and lower bounds only differ by 1/2 bit.

A. System under investigation

The bidirectional relay channel problem discussed above can be generalized to a multi-pair (or multiuser) setting in which the relay facilitates the communication between multiple pairs of users. The achievable degrees of freedom for a three user case with multiple antennas were determined in [14]. In [15] authors analyzed the case that the relay orthogonalizes different bidirectional transmissions by a distributed zero forcing algorithm and then multiple pairs communicate with each other via several orthogonalize-and-forward relay terminals. In [16], [17] authors investigated this problem for
interference limited systems in which each pair of users share a common spreading signature to distinguish themselves from the other pairs, and proposed a jointly demodulate-and-XOR forward strategy. However, so far no attempt has been done to characterize the capacity region of this network, and the optimal relaying strategy is unknown.

In this paper we study the information theoretic capacity of the multi-pair bidirectional wireless relay network. We first examine this problem in the context of the linear shift deterministic channel introduced by Avestimehr, Diggavi, and Tse [3]. This model simplifies the wireless network interaction model by eliminating the noise and allows us to focus on the interaction between signals. This approach was successfully applied to the relay network in [3], and resulted in insight in terms of transmission techniques which also led to an approximate characterization of the capacity of Gaussian relay networks. This approach has also been recently applied to the bidirectional relay channel problem [18], [19], which again resulted in finding near optimal relaying strategies as well as approximating the capacity region of the noisy (Gaussian) bidirectional relay channel. The deterministic approach is not restricted to relay networks. For instance, an approximate characterization of the capacity for the Gaussian interference channel was obtained in [20] using the deterministic approach. Transmission techniques in a deterministic relay-interference network were studied in [21].

B. Main contributions

Inspired by the results mentioned above, we apply the linear shift deterministic model to the multi-pair bidirectional relay network and analyze its capacity when the relay is operating at either full-duplex mode or half-duplex mode (with non-adaptive listen-transmit scheduling). In both cases we exactly characterize the capacity region and show that the cut-set upper-bound is tight. We show that the capacity region is achieved by dividing the signal level space elegantly between the multiple pairs, i.e., different pairs are orthogonalized on the signal level space. Each pair is then operating on the portion of the signal level space assigned to it. The relay uses a similar functional-forwarding scheme as in [18], in which the relay records the received superposed signals on the different levels and forwards them without decoding everything explicitly. The strategy is therefore referred to as divide-and-conquer-strategy.

Later on, we use these insights to find a near optimal transmission technique for the Gaussian case. More specifically, we propose a superposition of lattice codes and random Gaussian codes at the source nodes. However, orthogonalization as in the deterministic setup is not possible in the Gaussian setup as all signals arriving at the relay interact with each other. Thus the relay attempts to decode the Gaussian codewords of the respective nodes and the superposition of the lattice codewords of each pair by using successive interference cancelation. The relay then forwards this information to the intended destinations. We analyze the achievable rate region of this scheme and show that for all channel gains it achieves to within 3 bits/sec/Hz per user of the cut-set upper-bound on the capacity region of the two-pair bidirectional relay network.

The paper is organized as follows. In Section II we investigate the full-duplex and half-duplex multi-pair bidirectional linear shift deterministic relay network and characterize the exact capacity region of this network. In Section III we discuss the insights gained from the linear shift deterministic model and how these insights can be used in the Gaussian setup in the subsequent Section IV. In the Gaussian two-pair bidirectional relay network, we present upper bounds, derive our achievability strategy and characterize the constant gap between the upper bounds and our proposed scheme. We finally conclude the paper in Section V.

II. MULTI-PAIR BIDIRECTIONAL LINEAR SHIFT DETERMINISTIC RELAY NETWORK

In the following subsections, we state the precise definition of the problem and present the main result for the deterministic case.

A. System model

The system model for the \( M \)-pair bidirectional relay network is shown in Figure 1 In this system \( M \) pairs \((A_1, B_1), \ldots, (A_M, B_M)\) aim to use the relay to communicate with each other (i.e., \( A_1 \) and \( B_1 \) want to communicate with each other, and so on). The relay can operate on either full-duplex or half-duplex mode. In the full-duplex mode it is able to listen and transmit at the same time, while in the half-duplex mode it can only listen or transmit at a particular time. In the full-duplex scenario, we only consider the case that the listen-transmit scheduling is non-adaptive and the relay listens a fixed \( \Delta \) fraction of the time and transmits the rest. Although \( \Delta \) can not change adaptively as a function of the channel gains, one can optimize over \( \Delta \) beforehand.

We use the linear shift deterministic channel model to model the interaction between the transmitted signals. The linear shift deterministic channel model was introduced in [3]. Here is a formal definition of this channel model.

**Definition II.1:** (Definition of the linear shift deterministic model) Consider a wireless network as a set of nodes \( V \),
where $|V| = N$. Communication from node $i$ to node $j$ has a non-negative integer gain $n_{(i,j)}$ associated with it. This number models the channel gain in a corresponding Gaussian setting. At each time $t$, node $i$ transmits a vector $x_i[t] \in \mathbb{F}_2^q$ and receives a vector $y_i[t] \in \mathbb{F}_2^q$ where $q = \max_{i,j}(n_{(i,j)})$. The received signal at each node is a deterministic function of the transmitted signals at the other nodes, with the following input-output relation: if the nodes in the network transmit $x_1[t], x_2[t], \ldots, x_N[t]$ then the received signal at node $j$, $1 \leq j \leq N$ is:

$$y_j[t] = \sum_{k=1}^{N} S^{q-n_{k-j}} x_k[t]$$  

(1)

for all $1 \leq k \leq N$, where $S$ is the $q \times q$ shift matrix and the summation and multiplication is in $\mathbb{F}_2$.

Now that we have defined the linear shift deterministic channel model we can apply it to the multi-pair bidirectional relay network. A pictorial representation of an example of such network with two pairs is shown in Figure 2. In this figure each little circle represents a signal level and what is sent on it is a bit. The transmit and received signal levels are sorted from MSB to LSB from top to bottom. The channel gain between two nodes $i$ and $j$ indicates how many of the first MSB transmitted signal levels of node $i$ are received at destination node $j$. As described in the channel model (1), at each received signal level, the receiver gets only the modulo two summation of the incoming bits.

B. Cut-set upper-bound and a motivating example

The cut-set upper-bound [22] on the capacity region of the full-duplex $M$-pair bidirectional linear shift deterministic relay network (described in Section II-A) is given by

$$\sum_{i \in \mathcal{U}} [\ell_i R_{A_i} + (1 - \ell_i) R_{B_i}] \leq \min \left( \max_{i \in \mathcal{U}} (\ell_i n_{A_i R} + (1 - \ell_i) n_{B_i R}), \right.$$  

$$\max_{i \in \mathcal{U}} (\ell_i n_{R_{B_i}} + (1 - \ell_i) n_{R_{A_i}}) \right),$$(2)

for all $\mathcal{U} \subset \{1, \ldots, M\}$ and $\ell_i \in \{0, 1\}, i = 1, \ldots, M$. This bound is simply obtained by considering the pairs $(A_i, B_i)$, $i \in \mathcal{U}$, and creating a cut between them such that, if $\ell_i = 1$, $A_i$ is on the left and $B_i$ is on the right side of the cut, and if $\ell_i = 0$, $B_i$ is on the left and $A_i$ is on the right side of the cut. We then consider the sum-rate of communication from the nodes on the left side of the cut to the nodes on the right side of the cut. This is upper bounded by (2), where the first term on the RHS of (2) is the maximum number of bits that the relay can receive from the nodes on the left side of the cut, and the second term on the RHS of (2) is the maximum number of bits that the relay can broadcast to the nodes on the right side of the cut.

For example, in the case that we have only two pairs ($M = 2$) and the relay is operating on the full-duplex mode, the cut-set upper-bound on the capacity region is given by

$$R_{A_1} \leq \min (n_{A_1 R}, n_{R_{B_1}})$$  

(3)

$$R_{B_1} \leq \min (n_{B_1 R}, n_{R_{A_1}})$$  

(4)

$$R_{A_2} \leq \min (n_{A_2 R}, n_{R_{B_2}})$$  

(5)

$$R_{B_2} \leq \min (n_{B_2 R}, n_{R_{A_2}})$$  

(6)

$$R_{A_1} + R_{A_2} \leq \min (\max (n_{A_1 R}, n_{A_2 R}), \max (n_{R_{B_1}}, n_{R_{B_2}}))$$  

(7)

$$R_{B_1} + R_{B_2} \leq \min (\max (n_{B_1 R}, n_{B_2 R}), \max (n_{R_{A_1}}, n_{R_{A_2}}))$$  

(8)

$$R_{A_1} + R_{B_2} \leq \min (\max (n_{A_1 R}, n_{B_2 R}), \max (n_{R_{B_1}}, n_{R_{A_2}}))$$  

(9)

$$R_{B_1} + R_{A_2} \leq \min (\max (n_{B_1 R}, n_{A_2 R}), \max (n_{R_{A_1}}, n_{R_{B_2}})).$$  

(10)

As a motivating example, we now consider the network shown in Figure 2. It is easy to check that the rate tuple

$$(R_{A_1}, R_{B_1}, R_{A_2}, R_{B_2}) = (2, 1, 1, 1)$$

is inside its cut-set region. In Figure 3 we illustrate a simple scheme that achieves this rate point. With this strategy, the nodes in the uplink transmit

$$x_{A_1} = [a_{1,1}, a_{1,2}, 0]^t, \quad x_{B_1} = [b_{1,1}, 0, 0]^t$$  

$$x_{A_2} = [0, a_{2,1}, 0]^t, \quad x_{B_1} = [b_{2,1}, 0, 0]^t$$

and the relay receives

$$y_R = [a_{1,1}, a_{1,2} \oplus b_{1,1}, a_{2,1} \oplus b_{2,1}]^t.$$

Then the relay will re-order the received signal and transmit

$$x_R = [a_{2,1} \oplus b_{2,1}, a_{1,2} \oplus b_{1,1}, a_{1,1}]^t.$$

Then node $A_1$ receives the signals (i.e., XOR-combination) $a_{1,2} \oplus b_{1,1}$ and since it knows $a_{1,2}$ can decode $b_{1,1}$. Similarly node $B_1$ can decode $a_{1,1}$ and $a_{1,2}$, node $A_2$ can decode $b_{2,1}$ and finally node $B_2$ can decode $a_{2,1}$. Therefore we achieve the rate point $(2, 1, 1, 1)$.

There are some interesting points about this particular achievability strategy:

- There is no coding over time.
- There is no interference between different pairs on the same received signal level at the relay.
- The relay just re-orders the received XOR-combinations and forwards them.

1Some channels may have zero gain.
strategy the in Section II-A, is equal to the cut-set upper-bound (2), and it is

\[ R = \prod_{i=1}^{M} (R_{A_i} + R_{B_i}) \]

to show that every integral 2M-tuple \((R_{A_1}, R_{B_1}, \ldots, R_{A_M}, R_{B_M})\) satisfying the cut set bound is achievable by allocating subsets of the signal levels exclusively to users of different session\(^3\), and using functional-forwarding at the relay.

The proof is obvious for \(R_{\text{sum}} = 1\). Assume it is true for all channel gains and all integral rate-tuples with sum-rate \(R_{\text{sum}} \leq k\). We now prove this for \(R_{\text{sum}} = k + 1\). Consider a 2M-tuple \(R = (R_{A_1}, R_{B_1}, \ldots, R_{A_M}, R_{B_M})\) satisfying the cut set bound (2) and \(R_{\text{sum}} = k + 1\). We consider two separate cases.

Case 1: There is a pair where both nodes have nonzero transmission rates. Without loss of generality we may assume that \(R_{A_1}\) and \(R_{B_1}\) are both nonzero. Our goal is to choose one up-link signal level and one down-link signal level at the relay, and assign them to the \((A_1, B_1)\) session. \(A_1\) and \(B_1\) will then transmit one bit at the specified uplink level to the relay, and the relay will transmit (broadcast) the received XOR-combination at the specified down-link level to both \(A_1\) and \(B_1\). After doing so and removing the specified signal levels, the network will reduce to a network with lower channel gains. We then show that the reduced rate-tuple \((R_{A_1} - 1, R_{B_1} - 1, R_{A_2}, R_{B_2}, \ldots, R_{A_M}, R_{B_M})\) is in the cut-set region of the reduced network. Therefore, by induction, it will be achieved and the proof will be complete.

More specifically, for the up-link we choose the highest signal level connected to both \(A_1\) and \(B_1\) (denoted by \(l_u = \min(n_{A_1}, n_{B_1})\)), and for the down-link, we chose the lowest signal level connected to both \(A_1\) and \(B_1\) (denoted by \(l_d = \min(n_{RA_1}, n_{RB_1})\)). After removing signal levels \(l_u\) and \(l_d\) from the up-link and down-link of the relay, we obtain a linear shift deterministic network with channel gains

\[
\begin{align*}
    n'_{A_1R} &= n_{A_1R} - 1(n_{A_1R} \geq l_u), \quad i = 1, \ldots, M, \quad (11) \\
    n'_{B_1R} &= n_{B_1R} - 1(n_{B_1R} \geq l_u), \quad i = 1, \ldots, M, \quad (12) \\
    n'_{RA_i} &= n_{RA_i} - 1(n_{RA_i} \geq l_d), \quad i = 1, \ldots, M, \quad (13) \\
    n'_{RB_i} &= n_{RB_i} - 1(n_{RB_i} \geq l_d), \quad i = 1, \ldots, M, \quad (14)
\end{align*}
\]

where \(1(.)\) is the indicator function.

As we show in Appendix A the reduced rate-tuple \(R' = (R_{A_1} - 1, R_{B_1} - 1, R_{A_2}, R_{B_2}, \ldots, R_{A_M}, R_{B_M})\) is in the cut-set region of this network, i.e.,

\[
\sum_{i \in U} \ell_i (R'_{A_i} + (1 - \ell_i)R'_{B_i}) \leq \min \left( \max_{i \in U} (\ell_i n'_{A_iR} + (1 - \ell_i)n'_{B_iR}), \quad \max_{i \in U} (\ell_i n'_{RB_i} + (1 - \ell_i)n'_{RA_i}) \right),
\]

for all \(U \subseteq \{1, \ldots, M\}\) and \(\ell_i \in \{0, 1\}, i = 1, \ldots, M\)

Moreover, \(R'\), has sum-rate \(R'_{\text{sum}} = R_{\text{sum}} - 2 = k - 1\). Hence, by our induction assumption, it can be achieved by

We call a strategy with these properties a divide-and-conquer relaying strategy, which will be defined more formally in the next section. Quite interestingly, we next prove that any rate point in the cut-set bound region of the bidirectional linear shift deterministic relay network can be achieved using such strategy.

C. Capacity region

In this section we study the capacity region of the multi-pair bidirectional linear shift deterministic relay network. We first give an overview of our achievability strategy. This strategy consists of three components, namely orthogonalization, re-ordering (or permutation) and forwarding. The first component (orthogonalization) divides the uplink signal levels at the relay between the pairs in such a way that no signal-level is assigned to more than one-pair. Hence, different pairs are orthogonalized in the uplink and do not interact with each other anymore. As a consequence, at each level, the relay either receives one bit from a single node or the XOR-combination of two bits coming from the pair of nodes that wish to communicate with each other. The relay then reorders its received signal by applying a permutation matrix \(\Pi\) (i.e., \(x_R = \Pi y_R\)) and forwards it in the downlink. We name this strategy the divide-and-conquer relaying strategy.

We now state our main result in this section.

**Theorem 1**: The capacity region of the full-duplex multi-pair bidirectional linear shift deterministic relay network, described in Section II-A, is equal to the cut-set upper-bound (2), and it is achieved by the divide-and-conquer relaying strategy described above.

**Proof:**
We first prove the result for integral\(^2\) rate-tuples. We use

\[^2\text{i.e., with integer components.}\]

\[^3\text{A session means the communication of one pair.}\]
using the remaining levels and the proof in this case is complete.

*Case 2:* Every session has a node with zero rate. Without loss of generality, assume that \( R_{B_1} = \cdots = R_{B_M} = 0 \) and \( R_{A_1} \geq 1 \). Again, we choose one up-link signal level and one down-link signal level at the relay, and assign them to the \((A_1, B_1)\) session. \( A_1 \) will then transmit one bit at the specified uplink level to the relay, and the relay will transmit (broadcast) the received bit at the specified down-link level to \( B_1 \). After doing so and removing the specified signal levels, the network will reduce to a network with lower channel gains. We then show that the reduced rate-tuple \((R_{A_1}-1, 0, R_{A_2}, 0, \ldots, R_{A_M}, 0)\) is in the cut-set region of the reduced network. Therefore, by induction, it will be achieved and the proof will be complete.

More specifically, we choose the highest signal level in the up-link that is connected \( A_1 \) (denoted by \( l_u = n_{A,R} \)), and for the down-link, we chose the lowest signal level connected to \( B_1 \) (denoted by \( l_d = n_{RB_1} \)). After removing signal levels \( l_u \) and \( l_d \) from the up-link and down-link of the relay, we obtain a linear shift deterministic network with channel gains in \( \mathbf{11} - \mathbf{14} \).

As we show in Appendix \( \mathbf{15} \) the reduced rate-tuple \( R' = (R_{A_1}-1, 0, R_{A_2}, 0, \ldots, R_{A_M}, 0) \) is in the cut-set region of the reduced network. Moreover, it has sum-rate \( R'_{\text{sum}} = R_{\text{sum}} - 1 = k \). Hence, by our induction assumption, it can be achieved and the proof in this case is complete.

To complete the proof, we just need to show that all corner points of the cut-set bound region are achieved by the divide-and-conquer relaying strategy. Note that since all coefficients of the hyperplanes of the cut-set bound region are integers, if a fraction of the network over time is the same as the original network with all channel gains multiplied by a shift matrix. Hence, we know that the capacity region of this full-duplex network is equal to its cut-set upper-bound (17). Since the relay is operating on the half-duplex mode (i.e., listening a fraction of the time and transmitting the rest), the cut-set upper-bound (22) on the capacity region of \( M \)-pair bidirectional linear shift deterministic relay network will be

\[
\sum_{i \in \mathcal{U}} [\ell_i R_{A_i} + (1 - \ell_i) R_{B_i}] 
\leq \min \left( \Delta \max_{i \in \mathcal{U}} (\ell_i n_{A,R} + (1 - \ell_i) n_{B,R}), \right.
(1 - \Delta) \max_{i \in \mathcal{U}} (\ell_i n_{RB}, (1 - \ell_i) n_{R_{A_i}}),
\]

for all \( \mathcal{U} \subseteq \{1, \ldots, M\} \) and \( \ell_i \in \{0, 1\}, i = 1, \ldots, M \).

As a corollary of Theorem \( \mathbf{11} \) we can also show that in this case the cut-set upper-bound is achievable.

**Corollary 1:** Let \( \Delta \) be the fraction of the time the relay listens and transmits the rest. The capacity region of the multi-pair bidirectional linear shift deterministic relay network with a half-duplex relay is equal to the cut-set upper-bound (17), and it is achieved by the divide-and-conquer relaying strategy.

**Proof:** Without loss of generality assume \( \Delta \) is a fractional number (otherwise consider the sequence of fractional numbers approaching it). Then choose a large enough integer \( Q \) such that \( Q \Delta \) is integer. Then consider \( Q \) instances of the network over time, such that for \( Q \Delta \) instances the relay is listening and in the other \( Q(1 - \Delta) \) instances it is transmitting. After concatenating these instances together, the resulting network can be thought of as a full-duplex multi-pair network where the uplink channel gains are multiplied by \( Q \Delta \) and the downlink channel gains are multiplied by \( (1 - \Delta)Q \). It is easy to verify that the cut-set bound region of this network is just the cut-set bound region of the original half-duplex network expanded by \( Q \). Now by Theorem \( \mathbf{11} \) and the previous argument, we know that the capacity region of this full-duplex multi-pair bidirectional network is equal to its cut-set upper-bound and is achieved by the divide-and-conquer relaying strategy. Now note that any divide-and-conquer relaying strategy in this full-duplex network can be translated to a divide-and-conquer relaying strategy in \( Q \) instances of the original half-duplex network; \( Q \Delta \) instances the relay is in the listen mode to get the signals and \( (1 - \Delta)Q \) instances in the transmit mode to forward the signals. Therefore the cut-set upper-bound...
\[ \hat{x}_k = [x_k(1)[0], \ldots, x_k(1)[Q-1], x_k(2)[0], \ldots, x_k(2)[Q-1], \ldots, x_k(q)[0], \ldots, x_k(q)[Q-1]]^T \in \mathbb{F}_2^{Qq} \]
\[ \hat{y}_k = [y_k(1)[0], \ldots, y_k(1)[Q-1], y_k(2)[0], \ldots, y_k(2)[Q-1], \ldots, y_k(q)[0], \ldots, y_k(q)[Q-1]]^T \in \mathbb{F}_2^{Qq}. \] (16)

Fig. 4. Illustration of the inductive algorithm introduced in Theorem 1.
is achievable and the proof is complete.

D. Remark

An interesting insight, which will prove to be useful in the transition to the Gaussian case is the following. Although the scheme we provided is an inductive way of level assignment and seems quite unstructured (in the sense that it assigns signal levels on a greedy basis), one can actually say more about these assignments using certain observations. First of all, note that in this divide-and-conquer relaying strategy we have in general $2M$ types of signals that the relay might decode. Namely, $M$ types of signals that are made up of one bit from one user of a session, and $M$ types of signals that are the XOR-combination of bits from both users of the same pair. Each signal is received at the relay at some signal level, and is transmitted to one or both of the end users at potentially another signal level in down-link. Please refer to the example network of Figures 4 and 5 and observe that quite interestingly, in the final configuration of signal type-level assignments, all signals of the same type are concatenated together both in UL and DL. In other words they appear at concatenated signal levels. In general, one can serve all signals of the same type at once by choosing a pair with nonzero rates and serve them (one bit per user per signal level) until one of the rates is zero. For $M = 2$ for example, assuming $R_{A_1} \geq R_{B_1}$ and $R_{A_2} \geq R_{B_2}$, instead of reducing $(R_{A_1}, R_{B_1}, R_{A_2}, R_{B_2})$ to $(R_{A_1} - 1, R_{B_1} - 1, R_{A_2}, R_{B_2})$ one can reduce it to $(R_{A_1} - R_{B_1}, 0, R_{A_2}, R_{B_2})$ all at once and find a chunk of signal levels to afford them. Then the same thing can be done for the other pair. In the final configuration, all signals of the same type are in concatenation, which is also illustrated in Fig. 6.

In the following, we discuss more insights gained from the examination of the linear shift deterministic multi-pair bidirectional relay network that can be interpreted for the two-pair Gaussian relay network.

III. TRANSITION FROM THE LINEAR SHIFT DETERMINISTIC MODEL TO GAUSSIAN MODEL

The result of the deterministic network basically suggests that it is optimal to divide the signal-level space into subspaces and allocate these orthogonal subspaces to the different sessions, i.e., pairs. Furthermore, it suggests to split the message of the stronger user of each pair (the user with stronger uplink channel, cf. section II-D) into two parts:

1) the first part has the same rate as the message from the weak user and it is transmitted such that at the relay it is received with the same power as that of the signal from the weak user,

2) the second part has the remaining rate and is transmitted at some higher signal levels.

Hence, for $M = 2$, the relay receives four chunks of bits at different signal levels. Namely, the bits that are created from the XOR-combination of the signals of both users of each pair and, bits from the signals of the strong transmitter of each pair. The relay then forwards these signals at non-overlapping signal levels to the end users so that the XOR-combination of the signals is received by both users, whereas the other bits (from the strong transmitters) are received by the corresponding end users only. This way each user can easily decode its message having the received XOR-combinations, received bits and its own transmitted message.

To apply a similar strategy to Gaussian networks, one will face three immediate challenges. The first one is the effect of the additive noise which is inevitably present in the Gaussian channels. The second issue is that the received signals at the relay can not be fully orthogonalized (i.e., we face interference between low power and high power signals). The third complication is in decoding the superposition of signals (and not the individual signals) which should take place at the relay.

We propose the following solutions to overcome these difficulties. The noise issue can be simply resolved by using an appropriate block symbol coding scheme. The orthogonalization problem is inevitable, however a compensation in the capacity region allows for interference tolerance. In other words, rather than showing the cut-set upper-bound is tight, we show that the cut-set upper-bound is achievable to
within a constant. Finally, using an appropriate lattice code, the third challenge is resolvable, too. In a lattice structure, the superposition of every two codewords is also a lattice codeword and therefore can be decoded at the relay [11], [12]. These will be addressed in the sections that follow.

IV. TWO-PAIR BIDIRECTIONAL GAUSSIAN RELAY NETWORK

In this section we analyze the capacity region of the two-pair bidirectional Gaussian relay network shown in Figure 7. In particular, we show that the transmission scheme which was developed in the previous section achieves within 3 bits/sec/Hz per user of the cut-set upper-bound on the capacity region.

Thus, we consider two single-antenna transceiver pairs, \((A_1, B_1)\) and \((A_2, B_2)\), communicating to each other by exploiting a relay \(R\). The relay is operating in the full-duplex mode, i.e., it can listen and transmit at the same time. We use a complex AWGN channel model for all channels in this network. Hence, the received signals at the nodes are given by

\[
y_R = h_{A_1} x_{A_1} + h_{B_1} R x_{B_1} + h_{A_2} R x_{A_2} + h_{B_2} R x_{B_2} + z_R,
\]

\[
y_{A_i} = h_{R A_i} x_{R} + z_{A_i}, \quad y_{B_i} = h_{R B_i} x_{R} + z_{B_i}, \quad i = 1, 2
\]

where \(x_{A_1}, x_{B_1}, x_{A_2}, x_{B_2}\), and \(x_{R}\) are the signals transmitted from nodes \(A_1, B_1, A_2, B_2\), and \(R\), respectively. The transmit power constraint is \(E[|x_{A_1}|^2] = E[|x_{B_1}|^2] = E[|x_{R}|^2] \leq P\) and the noises \(z_{A_1}, z_{B_1}, z_{A_2}, z_{B_2}\), and \(z_R\) are all distributed as \(CN(0, 1)\).

Note that the uplink channels gains \(h_{A_i}\) and \(h_{B_i}\) are not necessarily equal to the down-link channel gains \((h_{R A_i}, h_{R B_i})\), i.e., channel reciprocity is not assumed. For each pair \((A_i, B_i)\), \(R_{A_i}\) is the rate at which \(A_i\) transmits data to \(B_i\) and \(R_{B_i}\) is the transmission rate of \(B_i\) to \(A_i\).

Now begin by describing the cut-set upper-bound [22], denoted by \(\bar{c}_{cs}\), on the capacity region of this network:

\[
\bar{c}_{cs} = \left\{ (R_{A_1}, R_{B_1}, R_{A_2}, R_{B_2}) \in \mathbb{R}^4_+: \right. \]

\[
R_{A_1} \leq \min \left( C \left( |h_{A_1} R|^2 P \right), C \left( |h_{R A_1}|^2 P \right) \right) \]  

\[
R_{B_i} \leq \min \left( C \left( |h_{B_i} R|^2 P \right), C \left( |h_{R B_i}|^2 P \right) \right) \]  

\[
R_{A_1} + R_{A_2} \leq \min \left( C \left( \left( |h_{A_1} R| + |h_{A_2} R| \right)^2 P \right), \right. \]

\[
C \left( \left( |h_{R A_1}|^2 + |h_{R A_2}|^2 P \right) \right) \]  

\[
R_{B_1} + R_{B_2} \leq \min \left( C \left( \left( |h_{B_1} R| + |h_{B_2} R| \right)^2 P \right), \right. \]

\[
C \left( \left( |h_{R B_1}|^2 + |h_{R B_2}|^2 P \right) \right) \]  

\[
R_{A_1} + R_{B_2} \leq \min \left( C \left( \left( |h_{A_1} R| + |h_{B_2} R| \right)^2 P \right), \right. \]

\[
C \left( \left( |h_{R B_1}|^2 + |h_{R B_2}|^2 P \right) \right) \]  

where \(C(x) = \log (1 + x)\). The terms in (26)-(31) correspond to the cuts labeled from 1 to 8 in Fig. 8.

We also define a ‘restricted cut-set bound’, denoted by \(\bar{c}\), to be:

\[
\bar{c} = \left\{ (R_{A_1}, R_{B_1}, R_{A_2}, R_{B_2}) \in \mathbb{R}^4_+: \right. \]

\[
R_{A_1} \leq \min \left( C \left( |h_{A_1} R|^2 P \right), C \left( |h_{R A_1}|^2 P \right) \right) \]  

\[
R_{B_i} \leq \min \left( C \left( |h_{B_i} R|^2 P \right), C \left( |h_{R B_i}|^2 P \right) \right) \]  

\[
R_{A_1} + R_{A_2} \leq \min \left( C \left( \left( |h_{A_1} R|^2 + |h_{A_2} R|^2 \right) P \right), \right. \]

\[
C \left( \max \left( |h_{R A_1}|^2, |h_{R A_2}|^2 \right) P \right) \]  

\[
R_{B_1} + R_{B_2} \leq \min \left( C \left( \left( |h_{B_1} R|^2 + |h_{B_2} R|^2 \right) P \right), \right. \]

\[
C \left( \max \left( |h_{R B_1}|^2, |h_{R B_2}|^2 \right) P \right) \]  

\[
R_{A_1} + R_{B_2} \leq \min \left( C \left( \left( |h_{A_1} R|^2 + |h_{A_2} R|^2 \right) P \right), \right. \]

\[
C \left( \max \left( |h_{R A_1}|^2, |h_{R A_2}|^2 \right) P \right) \]  

\[
R_{B_1} + R_{A_2} \leq \min \left( C \left( \left( |h_{B_1} R|^2 + |h_{A_2} R|^2 \right) P \right), \right. \]

\[
C \left( \max \left( |h_{R B_1}|^2, |h_{R B_2}|^2 \right) P \right) \} \]

Note the gap between the cut-set bound and the restricted cut-set bound is at-most 1 bit/sec/Hz per user.

Lemma 1: The cut-set upper bound in (18) is within 1 bit/sec/Hz per user of the restricted cut-set upper bound in (25).

Proof: Consider the first expressions in (21) and (28). It holds that

\[
C \left( \left( |h_{A_1} R|^2 + |h_{A_2} R|^2 \right) P \right) 
\leq C \left( \left( \left| |h_{A_1} R| \right|^2 + 2 |h_{A_1} R| |h_{A_2} R| \right| + \left| h_{A_2} R \right|^2 \right) P \right) 
\leq C \left( \left( \left( |h_{A_1} R|^2 + 2 |h_{A_2} R|^2 \right) P \right) 
\leq C \left( \left( \left( |h_{A_1} R|^2 + |h_{A_2} R|^2 \right) P \right) 
\right)
\]

where (a) follows since \( |h_{A_1} R| - |h_{A_2} R| | \geq 0 \). Thus, the gap between the first expressions in (21) and (28) is at most 1 bit/sec/Hz. Similarly, for the second expressions in (21) and (28), it holds that

\[
C \left( \left( |h_{R B_1}|^2 + |h_{R B_2}|^2 \right) P \right) 
\leq C \left( \left( \left( 2 |h_{A_1} R| + 2 |h_{A_2} R|^2 \right) P \right) 
\leq C \left( \left( \left( |h_{R B_1}|^2 + |h_{R B_2}|^2 \right) P \right) 
\right)
\]

and thus the gap between he second expressions in (21) and (25) is at most 1 bit/sec/Hz. Following the same procedure for the remaining sum rate terms in (18) and (25) completes the proof.

In the remainder of the paper, we only consider the restricted cut-set upper bound. This is motivated as follows. The structure of the expressions in (25) resemble the rate
expressions of the achievable scheme which is described in
the following. Thus, the gap analysis becomes very convenient
and by Lemma 1 we are assured that we loose at most one
additional bit/sec/Hz in the gap analysis to go from the
restricted cut-set bound to the actual cut-set bound.

Next, we define the up-link and down-link cut-set regions.
The up-link cut-set region, $C_u$, is the set of rates satisfying
equations (26)-(31) when the down-link channel gains are
assumed infinity. This means that the only restricting factors
in determining the capacity regions are assumed to be the
up-link channel gains. Likewise, the down-link cut-set region,
$C_d$, is the set of rates satisfying (26)-(31) in which the up-link
channel gains are set to infinity. Note that $\bar{C} = C_d \cap C_u$.

We say that a 4-tuple $(R_{A_i}, R_{B_i}, R_{A_2}, R_{B_2})$ is achievable
if simultaneously $A_i$ can communicate to $B_i$ at rate $R_{A_i}$
and $B_i$ can communicate to $A_i$ at rate $R_{B_i}$ with arbitrary small
error probability. The union of all achievable rate tuples is
defined as the capacity region. We are now ready to state our
main result.

**Theorem 2:** The capacity region of the two pair full-duplex
bidirectional relay network is within 2 bits/sec/Hz per user of
its restricted cut-set upper-bound described in (26)-(31). Or,
more precisely, if

$$(R_{A_1}, R_{B_1}, R_{A_2}, R_{B_2}) \in \bar{C}$$

and $R_{A_i}, R_{B_i} \geq 2$ for $i = 1, 2$, then the rate tuple $(R_{A_1} - 2, R_{B_1} - 2, R_{A_2} - 2, R_{B_2} - 2)$ is achievable.

The rest of this section is devoted to proving this Theorem.
First, we state the following lemma which helps us by limiting
the number of rate configurations that we have to consider.

**Lemma 2:** Let \( R = (R_{A_1}, R_{B_1}, R_{A_2}, R_{B_2}) \) be a rate tuple
in the cut-set region $\bar{C}$. Assume $R_{A_i} \geq R_{B_i}$, $i = 1, 2$. Then
it is always possible to sufficiently reduce the transmit powers
at the uplink and add extra noise to the received signals at
the downlink, such that new effective channel gains satisfy
$|h_{A_i,R_i}| \geq |h_{B_i,R_i}|$ and $|h_{R_i,B_i}| \geq |h_{R_i,A_i}|$ for $i = 1, 2$, and \( R \) is
still in the shrunk cut-set region.

**Proof:** See Appendix C. This lemma basically reduces the number of relevant channel
gain orderings that we have to consider in order to prove
Theorem 2. Assume that the rate tuple that we want to show to be achievable (within 2 bits/sec/Hz per user) satisfies
$R_{A_i} \geq R_{B_i}$ for $i = 1, 2$. By Lemma 2, we can without loss of generality (wlog) assume that $|h_{A_i,R_i}| \geq |h_{B_i,R_i}|$ for
$i = 1, 2$. We can also wlog assume that $|h_{A_1,R_1}| \geq |h_{B_2,R_2}|$
(otherwise we can de-label pair 1 and pair 2). Therefore, we
only need to consider three different channel gain orderings
for the uplink. Those three cases are shown in Fig. 8(a) and
8(b). Similarly, we only need to consider three cases for the
downlink. To prove Theorem 2, we first describe the
encoding strategy at the transmission nodes. As mentioned
earlier, the idea is that strong transmitters of each pair split
their signals into a Gaussian codeword and a lattice codeword,
while the weak user only transmits a lattice codeword. While
stating this encoding strategy we leave the power allocation
parameters unspecified. In other words, the power level at
which the user breaks up its message into the superposition
of Gaussian and a lattice codeword remains as parameters.
In the next step we mention the decoding at the relay where the
superposition of lattice points and the Gaussian codewords
are decoded. Afterwards, the relay maps each of the four
decoded codewords into a random Gaussian codeword, and
broadcasts their weighted superposition to all users. The last
step is the decoding at the nodes, where every receiver first
decodes the undesired codewords that have larger weights than
the desired codewords. Thus, those codewords are decoded
and successively canceled from the received signal one by
one. Afterwards, both the weak and the strong receivers of
each pair decode the Gaussian codeword corresponding to the
lattice codeword belonging to that pair. In addition to that, the
strong receivers decode one more codeword. This codeword
corresponds to the Gaussian codeword, which was received by
the relay from their transmitting strong counterpart. Eventually
as a result of this scheme the rates that the users will suc-
cessfully transmit will be a function of the power parameters
that we set at the beginning. We will finally show that by
choosing these parameters appropriately any rate tuple within
2 bits/sec/Hz per user of the cut set is achievable.

### A. Lattice Coding

In the following, some preliminaries and results on lattice
coding are provided that we use in the remainder of the paper.
We refer the interested reader to 23 for more details.

A lattice $\Lambda$ of dimension $n$ is described by

$$\Lambda = \{ \lambda = Gx : x \in \mathbb{Z}^n \},$$

where $G$ describes the lattice and is referred to as the generator
matrix. The fundamental Voronoi region of such a lattice $A$
is denoted by $\Omega$. Furthermore, the volume of $\Omega$, i.e., the
reciprocal of the number of lattice point per unit volume,
is denoted by $V$. Now, let $p$ a positive integer and $Z_p$ the
set of integers modulo $p$. Further, let $\bar{v} : \mathbb{Z}^n \rightarrow \mathbb{Z}_p^n$ be the
componentwise modulo operation over integer vectors. The
lattices used in this paper are mod-$p$ lattices, i.e., of the form

$$\Lambda_c = \{ v \in \mathbb{Z}^n : \bar{v} \in C \},$$

where $C$ is a linear $(n,k)$ code over $Z_p$ and $p$ is prime 23
Construction A]. Now, let $B$ be a balanced set 23, 24 of
linear $(n,k)$ codes over $Z_p$ and let $\mathcal{L}_B$ be the set of lattices
denoted by

$$\mathcal{L}_B = \{ \Lambda_c : C \in B \};$$
the average probability of error with lattice decoding can be

Similarly, it was shown in [23] that by using a codebook

Therefore, we can assume

Then, the transmit signals at the nodes are given by

With this in mind, let’s consider the following system model

where \( y = x + z \),

\[ y = x + z, \]

where \( y \) is the receive signal, \( x \) is the transmit signal and \( z \) is additive noise with zero mean and a variance \( \sigma^2 \). It was shown in [23] that if the transmitted codeword is a lattice point, then there exists a lattice for that channel and the average probability of error with lattice decoding can be made arbitrarily small as the dimension of the lattice increases.

Similarly, it was shown in [23] that by using a codebook \( (A + s) \cap S \), where \( s \) is a shift and \( S \) describes the shaping gain, a rate \( R \) with arbitrarily small probability of error can be achieved if

\[ R \leq \log \left( \frac{P}{\sigma^2} \right). \]

We will use this result in the remainder of the paper for the characterization of the rate region achievable with our proposed scheme.

B. Encoding at the nodes

Wlog assume that \( R_{A,R} \geq R_{B,R} \). By Lemma 2 this means that we can assume \( |h_{A,R}| \geq |h_{B,R}| \) and \( |h_{R,B_i}| \geq |h_{R_A}| \). Then, the transmit signals at the nodes are given by

\[ x_{A_i} = \sqrt{\alpha^{(1)}_{A_i} x^{(1)}_{A_i}} + \sqrt{\alpha^{(2)}_{A_i} x^{(2)}_{A_i}}, \]

\[ x_B = \sqrt{\alpha^{(2)}_{B_1} x^{(2)}_{B_1}} \quad i = 1, 2 \]

\[ x = \sum_{j=1}^{4} \sqrt{\alpha^{(j)}_R x^{(j)}_R} \quad \text{with} \quad \sum_{j} \alpha^{(j)}_R = 1, \quad (32) \]

where \( x^{(1)}_{A_i} \) and \( x^{(2)}_{R} \) are codewords chosen from a random Gaussian codebook of size \( 2^{nR_{A_i}^{(1)}} \), \( i = 1, 2 \), and \( 2^{nR_{R}^{(2)}} \) for \( j = 1, \ldots, 4 \), respectively. \( x^{(2)}_{A_i} \) and \( x^{(2)}_{B_i} \), \( i = 1, 2 \), are lattice coded [23] using lattice ensembles \( \{ \Lambda^{(2)}_{A_i}, \Lambda^{(2)}_{A_2}, \Lambda^{(2)}_{B_1}, \Lambda^{(2)}_{B_2} \} \) giving a codebook of size \( 2^{nR_{A_i}^{(2)}} \) and \( 2^{nR_{R}^{(2)}} \) with \( i = 1, 2 \), respectively. We assume that the second moment per dimension of the fundamental Voronoi region [23] of each lattice is \( \frac{1}{2} \) which ensures satisfying the power constraint. At nodes \( A_i \) we have two messages \( m_{A_i}^{(1)} \) and \( m_{A_i}^{(2)} \) from dictionaries of size \( 2^{nR_{A_i}^{(1)}} \) and \( 2^{nR_{A_i}^{(2)}} \) that are mapped to \( x^{(1)}_{A_i} \) and \( x^{(2)}_{A_i} \), respectively. In other words, the strong transmitter of each pair transmits a superposition of a lattice code and a random Gaussian code, while the weaker user only transmits a lattice code. Thus, the transmit signals of nodes \( B_1 \) and \( B_2 \) reduce to

\[ x_{B_1} = \sqrt{\alpha^{(2)}_{B_1} x^{(2)}_{B_1}}, \]

\[ x_{B_2} = \sqrt{\alpha^{(2)}_{B_2} x^{(2)}_{B_2}}. \]

For the nodes \( A_1 \) and \( A_2 \), we have a superposition code (cf. (32)). Note that

\[ t = x^{(2)}_{A_1} + x^{(2)}_{B_1}, \]

\[ f = x^{(2)}_{A_2} + x^{(2)}_{B_2}, \]

where \( t \) and \( f \) are also lattice points due to the group structure of the lattice [12].

The power parameters (i.e., \( \alpha_{A_1} \) and \( \alpha_{B_1} \)) are assigned such that the lattice codes of each pair arrive at the same power level, so that the relay can decode the sum codeword correctly.

Thus we set,

\[ \alpha^{(2)}_{A_2} = \frac{|h_{B_1,R}|^2}{|h_{A,R}|^2} \alpha^{(2)}_{B_1}. \quad (33) \]

Furthermore, we should have \( \alpha^{(1)}_{A_1} + \alpha^{(2)}_{A_1} \leq 1 \) and \( \alpha^{(2)}_{B_1} \leq 1 \).

C. Uplink: Decoding at the relay

Recall that as discussed in Section III and illustrated in Figure 9 we have to analyze three cases only. Here, the analysis for the first case (cf. Fig. 9(a)) is given in detail. For the other cases, only the results are presented, since the other cases are similar and therefore omitted. However, along the presentation of the results, we also mention the differences should there be any.

1) Case \( |h_{A_1,R}| \geq |h_{B_1,R}| \geq |h_{A_2,R}| \geq |h_{B_2,R}| \).

The decoding order at the relay is as follows. First the relay decodes the Gaussian \( x^{(1)}_{A_1} \), then the lattice point \( t \) from \( A_1 \) and \( B_1 \), followed by \( x^{(2)}_{A_2} \) and finally the lattice point \( f \) from \( A_2 \) and \( B_2 \). We can show that for any choice of \( \alpha^{(1)}_{A_i} \) and \( \alpha^{(2)}_{B_i} \), this can be done successfully as long as,

\[ R_{A_i}^{(1)} \leq \frac{|h_{A_i,R}|^2 \alpha^{(1)}_{A_i}}{2 \alpha^{(2)}_{B_i} |h_{B_i,R}|^2 P + \alpha^{(1)}_{A_i} |h_{A_i,R}|^2 P + 2 \alpha^{(2)}_{B_i} |h_{B_i,R}|^2 P + 1} \]

(34)

Figure 9. Three relevant configurations for the uplink and their corresponding received signal at the relay. At the lowest level, all signals are superposed, while at the next level (medium shade), all but one signals are superposed. At the top level (white) only one signal remains.
\[ R_{A_i}^{(2)}, R_{B_i} \leq \log \left( \frac{|h_{B_i} R_i|^2 \alpha_{A_i}^{(2)} P}{\alpha_{A_i}^{(1)} |h_{A_i} R_i|^2 P + 2 \alpha_{B_i} |h_{B_i} R_i|^2 P + 1} \right) \]  
\[ R_{A_2}^{(2)}, R_{B_2} \leq \left( \log \left( \frac{\alpha_{B_2} |h_{B_2} R_2|^2 P}{2 \alpha_{B_2} |h_{B_2} R_2|^2 P + 1} \right) \right)^+ , \]  
\[ R_{A_2}^{(1)} \leq C \left( \frac{|h_{A_2} R_2|^2 \alpha_{A_2}^{(1)} P}{2 |h_{B_2} R_2|^2 \alpha_{A_2}^{(2)} P + 1} \right) . \]

Details of the derivations are given in Appendix [D].

2) Case \(|h_{A_1} R_1| \geq |h_{A_2} R_1| \geq |h_{B_1} R_1| \geq |h_{B_2} R_1|\): The decoding order at the relay is as follows. First the relay decodes the Gaussian \(x_A^{(1)}\) and \(x_A^{(2)}\) simultaneously by treating the remaining signals as noise. Afterwards, the lattice point \(t\) from \(A_1\) and \(B_1\) is decoded, followed by the lattice point \(f\) from \(A_2\) and \(B_2\). We can show that for any choice of \(\alpha_{A_1}^{(j)}\) and \(\alpha_{A_2}^{(2)}\), this can be done successfully as long as,

\[ R_{A_1}^{(1)} \leq C \left( \frac{|h_{A_1} R_1|^2 \alpha_{A_1}^{(1)} P}{2 \alpha_{B_1} |h_{B_1} R_1|^2 P + 2 \alpha_{B_2} |h_{B_1} R_1|^2 P + 1} \right) \]  
\[ R_{A_2}^{(2)} \leq \log \left( \frac{|h_{B_1} R_1|^2 \alpha_{A_1}^{(1)} P}{2 \alpha_{B_2} |h_{B_1} R_1|^2 P + 1} \right) \]  
\[ R_{A_2}^{(1)} \leq C \left( \frac{|h_{A_2} R_2|^2 \alpha_{A_2}^{(1)} P}{2 |h_{B_2} R_2|^2 \alpha_{A_2}^{(2)} P + 1} \right) . \]

There exists a choice of power assignments \(\alpha_{A_1}^{(j)}\) and \(\alpha_{A_2}^{(2)}\) such that the relay can use the decoding strategy described earlier to decode the Gaussian \(x_A^{(1)}\) of rate \(R_{A_1}^{(1)} = r_{A_1} - r_{B_1}\), the lattice point \(t\) of rate \(R_{A_2}^{(2)} = R_{B_1} = r_{B_1}\), and the lattice point \(f\) of rate \(R_{A_2}^{(2)} = R_{B_2} = r_{B_2}\), with arbitrary small error probability.

D. Encoding at the relay

The relay maps the decoded \(x_A^{(1)}\), \(t\), \(x_A^{(2)}\), and \(f\) to a Gaussian codeword \(x_R^{(1)}\) from a codebook of size \(2^{nR_{A_1}}\), \(x_R^{(2)}\) from a codebook of size \(2^{nR_{B_1}}\), \(x_R^{(3)}\) from a codebook of size \(2^{nR_{A_2}}\), and \(x_R^{(4)}\) from a codebook of size \(2^{nR_{B_2}}\), respectively.

E. Downlink: Decoding at the nodes

As in the uplink, we have to consider three cases only, from which we provide the detailed analysis for \(|h_{RB_1}| \geq |h_{RA_1}| \geq |h_{RB_2}| \geq |h_{RA_2}|\). The other cases follow similar lines of arguments and thus only the results are presented.

The relay uses a superposition of four messages. One message is decoded by all users. Another message is decoded by both users of the first pair and the strong receiver of the second pair. Yet another message is decoded by only the strong receiver of the first pair, and finally the remaining message is decoded by both users of the first pair.

1) Case \(|h_{RB_1}| \geq |h_{RA_1}| \geq |h_{RB_2}| \geq |h_{RA_2}|\): We can show that for any choice of \(\alpha_{A_1}^{(j)}\) and \(\alpha_{B_1}^{(2)}\), this can be done successfully as long as,

\[ R_{A_2}^{(1)}, R_{B_1} \leq \min \left( C \left( \frac{|h_{RB_1}|^2 \alpha_{A_1}^{(2)} P}{1 + |h_{RB_1}|^2 \alpha_{A_1}^{(1)} P} \right) , C \left( \frac{|h_{RA_1}|^2 \alpha_{B_1}^{(2)} P}{1 + |h_{RA_1}|^2 \alpha_{B_1}^{(1)} P} \right) \right) \]  
\[ R_{A_2}^{(2)}, R_{B_2} \leq \min \left( C \left( \frac{|h_{RB_2}|^2 \alpha_{B_2}^{(4)} P}{1 + P |h_{RB_2}|^2 \sum_{j=1}^{3} \alpha_{B_2}^{(j)} P} \right) , C \left( \frac{|h_{RA_2}|^2 \alpha_{B_2}^{(4)} P}{1 + P |h_{RA_2}|^2 \alpha_{B_2}^{(1)} + \alpha_{B_2}^{(2)} P} \right) \right) . \]
\[ R_{A_1}^{(1)} \leq C \left( |h_{RB_1}|^2 \alpha_R^{(1)} P \right), \]
\[ R_{A_2}^{(1)} \leq C \left( \frac{|h_{RB_2}|^2 \alpha_R^{(3)} P}{1 + P|h_{RB_2}|^2 \left( \alpha_R^{(1)} + \alpha_R^{(2)} \right)} \right), \]

Details of the derivation are given in Appendix \[ \square \]

2) Case \(|h_{RB_1}| \geq |h_{RB_2}| \geq |h_{RA_1}| \geq |h_{RA_2}|\): We can show that for any choice of \(\alpha_R^{(j)}\) and \(\alpha_B^{(n)}\), this can be done successfully as long as,
\[ R_{A_1}^{(2)}, R_{B_1} \leq \min \left( C \left( \frac{|h_{RA_1}|^2 \alpha_R^{(2)} P}{1 + P|h_{RA_1}|^2 \left( \alpha_R^{(1)} + \alpha_R^{(2)} \right)} \right), \right. \]
\[ C \left( \frac{|h_{RA_2}|^2 \alpha_R^{(2)} P}{1 + P|h_{RA_2}|^2 \left( \alpha_R^{(1)} + \alpha_R^{(2)} \right)} \right) \]

\[ R_{A_2}^{(2)}, R_{B_2} \leq \min \left( C \left( \frac{|h_{RB_1}|^2 \alpha_R^{(2)} P}{1 + P|h_{RB_1}|^2 \left( \alpha_R^{(1)} + \alpha_R^{(2)} \right)} \right), \right. \]
\[ C \left( \frac{|h_{RB_2}|^2 \alpha_R^{(2)} P}{1 + P|h_{RB_2}|^2 \left( \alpha_R^{(1)} + \alpha_R^{(2)} \right)} \right) \]

3) Case \(|h_{RB_1}| \geq |h_{RB_2}| \geq |h_{RA_2}| \geq |h_{RA_1}|\): We can show that for any choice of \(\alpha_R^{(j)}\) and \(\alpha_B^{(n)}\), this can be done successfully as long as
\[ R_{A_1}^{(2)}, R_{B_1} \leq \min \left( C \left( \frac{|h_{RB_1}|^2 \alpha_R^{(2)} P}{1 + P|h_{RB_1}|^2 \left( \alpha_R^{(1)} + \alpha_R^{(2)} \right)} \right), \right. \]
\[ C \left( \frac{|h_{RA_2}|^2 \alpha_R^{(2)} P}{1 + P|h_{RA_2}|^2 \left( \alpha_R^{(1)} + \alpha_R^{(2)} \right)} \right) \]

Now we state the following lemma whose proof is given in Appendix \[ \square \]

**Lemma 4:** Suppose that the relay is using the transmit strategy described above. Then for any 4-tuple \((r_{A_1}, r_{B_1}, r_{A_2}, r_{B_2})\) satisfying
\[ r_{A_1} \leq C \left( |h_{RB_1}|^2 P \right), r_{B_1} \leq C \left( |h_{RA_1}|^2 P \right) - 2 \]
\[ r_{A_2} \leq C \left( |h_{RB_2}|^2 P \right), r_{B_2} \leq C \left( |h_{RA_2}|^2 P \right) - 2 \]
\[ r_{A_1} + r_{A_2} \leq C \left( \max \left( |h_{RB_1}|^2 P, |h_{RB_2}|^2 P \right) \right) - 3 \]
\[ r_{A_1} + r_{B_2} \leq C \left( \max \left( |h_{RA_1}|^2 P, |h_{RA_2}|^2 P \right) \right) - 3 \]
\[ r_{B_1} + r_{A_2} \leq C \left( \max \left( |h_{RA_1}|^2 P, |h_{RA_2}|^2 P \right) \right) - 3 \]
\[ r_{B_1} + r_{B_2} \leq C \left( \max \left( |h_{RA_1}|^2 P, |h_{RA_2}|^2 P \right) \right) - 3 \]
there exists a choice of power assignments \(\alpha_R^{(j)}\)'s such that \(B_1\) can decode the Gaussian codewords \(x_R^{(1)}\) of rate \(R_{A_1}^{(1)} = r_{A_1} - r_{B_1}\), \(A_1\) and \(B_1\) can both decode the Gaussian codeword \(x_R^{(2)}\) of rate \(R_{A_2}^{(2)} = r_{B_1} - r_{B_2}\), \(B_2\) can decode the Gaussian codeword \(x_R^{(3)}\) of rate \(R_{A_1}^{(2)} = r_{A_2} - r_{B_2}\), and \(A_2\) and \(B_2\) can both decode the Gaussian codeword \(x_R^{(4)}\) of rate \(R_{A_2}^{(2)} = r_{B_2}\), with arbitrary small error probability.

Now note that if \((R_{A_1}, R_{B_1}, R_{A_2}, R_{B_2}) \in \overline{C}\) and \(R_{A_i}, R_{B_i} \geq 2\) for \(i = 1, 2\), then the rate tuple \((r_{A_1}, r_{B_1}, r_{A_2}, r_{B_2}) = (R_{A_1} - 2, R_{B_1} - 2, R_{A_2} - 2, R_{B_2} - 2)\) satisfies the conditions of both Lemma \[ \square \] and \[ \square \]. Therefore by the proposed strategy the rate tuple \((R_{A_1} - 2, R_{B_1} - 2, R_{A_2} - 2, R_{B_2} - 2)\) is achievable, and this completes the proof of Theorem \[ \square \]

V. Conclusion

In this paper we studied the multi-pair bidirectional relay network which is a generalization of the bidirectional relay channel. We examined this problem in the context of the linear shift deterministic channel model introduced in \[ \square \] and characterized its capacity region completely in both full-duplex and half-duplex cases. We also showed that the capacity can be achieved by a divide-and-conquer relaying strategy. Based on insights gained from the linear shift deterministic channel model, we proposed a transmission strategy for the Gaussian two-pair bidirectional full-duplex relay network and found an approximate characterization of the capacity region. In fact, we proposed a specific superposition coding scheme that achieves to within 3 bits/sec/Hz per user of the cut-set upper-bound on the capacity of the two-pair bidirectional relay network. Possible directions for future work is the extension to the half-duplex mode. Extension of the proposed transmission strategy to the case that there are more than two pairs is possible, however, analyzing the gap between the achievable rate of the corresponding scheme and the cut-set upper-bound is expected to be quite cumbersome.
APPENDIX A

In this Appendix we prove that the reduced rate-tuple \( \mathbf{R}' = (R_{A_1}-1, R_{B_1}-1, R_{A_2}, R_{B_2}, \ldots, R_{A_M}, R_{B_M}) \), with \( R_{A_1} \geq 1 \) and \( R_{B_1} \geq 1 \), created in case 1 of the proof of Theorem 1, is in the cut-set region of the reduced network (defined in (11)-(14)), i.e.,

\[
\sum_{i \in \mathcal{U}} \left[ \ell_i R_{A_i} + (1 - \ell_i) R_{B_i} \right] \\
\leq \min \left( \max_{i \in \mathcal{U}} (\ell_i n'_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \right. \\
\max_{i \in \mathcal{U}} (\ell_i n'_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}}) \right),
\]

for all \( \mathcal{U} \subseteq \{1, 2, \ldots, M\} \) and \( \ell_i \in \{0, 1\}, i = 1, 2, \ldots, M \).

If \( 1 \in \mathcal{U} \), we have

\[
\sum_{i \in \mathcal{U}} \left[ \ell_i R_{A_i} + (1 - \ell_i) R_{B_i} \right] - 1 \\
\leq \min \left( \max_{i \in \mathcal{U}} (\ell_i n'_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \right. \\
\max_{i \in \mathcal{U}} (\ell_i n'_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}}) \right)
\]

If \( 1 \notin \mathcal{U} \), we have

\[
\sum_{i \in \mathcal{U}} \left[ \ell_i R_{A_i} + (1 - \ell_i) R_{B_i} \right] - 1 \\
\leq \min \left( \max_{i \in \mathcal{U}} (\ell_i n'_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \right. \\
\max_{i \in \mathcal{U}} (\ell_i n'_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}}) \right) + 1,
\]

Therefore, if \( 1 \notin \mathcal{U} \), the only way to violate the cut-set bound is to have all above inequalities as equality, i.e.,

\[
\sum_{i \in \mathcal{U}} \left[ \ell_i R_{A_i} + (1 - \ell_i) R_{B_i} \right] = \min \left( \max_{i \in \mathcal{U}} (\ell_i n'_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \right. \\
\max_{i \in \mathcal{U}} (\ell_i n'_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}}) \right).
\]

However, we show that this is in contradiction to our assumption of \( R_{A_1} \neq 0 \) and \( R_{B_1} \neq 0 \). To see this, note that by (11)-(14), the equality in (70) happens only if we have one of the following four cases.

1) \( \exists j \in \{2, 3, \ldots, M\} \) such that, \( j \in \mathcal{U}, \ell_j = 1 \), and

\[
n'_{A_j, R} = \max_{i \in \mathcal{U}} (\ell_i n'_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \quad n'_{B_j, R} = \min_{i \in \mathcal{U}} \left( \max_{i \in \mathcal{U}} (\ell_i n'_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \right. \\
\max_{i \in \mathcal{U}} (\ell_i n'_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}}) \right),
\]

(71)

(72)

(73)

First, note that from (71), (72), \( \ell_j = 1 \), and the relationship between the channel gains in the original network and the channel gains in the reduced network (11)-(14), we have

\[
n_{A_j, R} = \max_{i \in \mathcal{U}} (\ell_i n_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \quad n_{B_j, R} = \min_{i \in \mathcal{U}} \left( \max_{i \in \mathcal{U}} (\ell_i n_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \right. \\
\max_{i \in \mathcal{U}} (\ell_i n_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}}) \right)
\]

(74)

(75)

(76)

Since \( n'_{A_j, R} = n_{A_j, R} - 1 \), we have \( n_{A_j, R} \geq \ell_u = \min(n_{A_j, R}, n_{B_j, R}) \). If \( n_{A_j, R} \geq n_{A_j, R} \), we can write

\[
R_{A_1} + \sum_{i \in \mathcal{U}} \left[ \ell_i R_{A_i} + (1 - \ell_i) R_{B_i} \right] \\
\leq \min \left( \max_{i \in \mathcal{U}} \left( n_{A_i, R}, \max_{i \in \mathcal{U}} (\ell_i n_{A_i, R} + (1 - \ell_i) n_{B_i, R}), \right. \right.
\max_{i \in \mathcal{U}} (n_{R_{A_i}}, \max_{i \in \mathcal{U}} (\ell_i n_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}})) \right)
\]

(77)

\[
\min \left( \max_{i \in \mathcal{U}} (n_{A_i, R}, n_{B_i, R}), \right. \right.
\max_{i \in \mathcal{U}} (n_{R_{A_i}}, \max_{i \in \mathcal{U}} (\ell_i n_{R_{A_i}} + (1 - \ell_i) n_{R_{B_i}})) \right)
\]

(78)

where the first step is true since \( \mathbf{R} \) satisfies the cut-set bound (2) with \( \mathcal{U} = \mathcal{U} \cup \{1\} \) and \( \ell_1 = 1 \). Combining (68) and (77), we get \( R_{A_1} \leq 0 \), which is a contradiction to our assumption of \( R_{A_1} \geq 1 \).
Similarly, if \( n_{A,R} \geq n_{B,R} \), we can write

\[
R_{B_i} + \sum_{i \in \mathcal{U}} [\ell_i R_{A_i} + (1 - \ell_i) R_{B_i}] 
\leq \min \left( \max \left( n_{B,R}, \max_{i \in \mathcal{U}} (\ell_i n_{A,R} + (1 - \ell_i) n_{B,R}) \right), \right. 
\left. \max \left( n_{R_A}, \max_{i \in \mathcal{U}} (\ell_i n_{R_{B_i}} + (1 - \ell_i) n_{R_A}) \right) \right)
\]

(84)

where the first step is true since \( R \) satisfies the cut-set bound with \( \mathcal{U} = \mathcal{U} \cup \{1\} \) and \( \ell_1 = 0 \). Combining (68) and (79), we get \( R_{B_i} \leq 0 \), which is a contradiction to our assumption of \( R_{B_i} \geq 1 \). Therefore, this case can not happen.

2) \( \exists j \in \{2, \ldots, M\} \) such that, \( j \in \mathcal{U}, \ell_j = 0 \), and

\[
n'_{B,R} = \max_{i \in \mathcal{U}} (\ell_i n'_{A,R} + (1 - \ell_i) n'_{B,R}),
n'_{R,B} = \min \left( \max_{i \in \mathcal{U}} (\ell_i n'_{A,R} + (1 - \ell_i) n'_{B,R}), \right. 
\left. \max_{i \in \mathcal{U}} (\ell_i n'_{R_{B_i}} + (1 - \ell_i) n'_{R_A}) \right)
\]

(85)

\( n'_{B,R} = n_{B,R} - 1 \). The proof that this case can not also happen is very similar to the previous case, hence we omit repetition.

3) \( \exists j \in \{2, \ldots, M\} \) such that, \( j \in \mathcal{U}, \ell_j = 0 \), and

\[
n'_{R,B_j} = \max_{i \in \mathcal{U}} (\ell_i n'_{R_{B_i}} + (1 - \ell_i) n'_{R_A}),
n'_{R,B} = \min \left( \max_{i \in \mathcal{U}} (\ell_i n'_{A,R} + (1 - \ell_i) n'_{B,R}), \right. 
\left. \max_{i \in \mathcal{U}} (\ell_i n'_{R_{B_i}} + (1 - \ell_i) n'_{R_A}) \right)
\]

(86)

\( n'_{R,B_j} = n_{R,B_j} - 1 \).

From (84), (85), \( \ell_j = 0 \), and the relationship between the channel gains in the reduced network (11)-(14), we have

\[
n_{R,B} = \max_{i \in \mathcal{U}} (\ell_i n_{R_{B_i}} + (1 - \ell_i) n_{R_A}),
\]

(83)

\[
n_{R,B} \leq \max_{i \in \mathcal{U}} (\ell_i n_{A,R} + (1 - \ell_i) n_{B,R}),
\]

(84)

\[
n_{R,B} = \min \left( \max_{i \in \mathcal{U}} (\ell_i n_{A,R} + (1 - \ell_i) n_{B,R}), \right. 
\left. \max_{i \in \mathcal{U}} (\ell_i n_{R_{B_i}} + (1 - \ell_i) n_{R_A}) \right).
\]

(85)

Since \( n'_{R,B} = n_{R,B} - 1 \), we have \( n_{R,B} \geq n_{R,B} - 1 \). If \( n_{R,B} \geq n_{R,B} \), we can write

\[
R_{A_i} + \sum_{i \in \mathcal{U}} [\ell_i R_{A_i} + (1 - \ell_i) R_{B_i}] 
\leq \min \left( \max_{i \in \mathcal{U}} (\ell_i n_{A,R} + (1 - \ell_i) n_{B,R}), \right. 
\left. \max_{i \in \mathcal{U}} (\ell_i n_{R_{B_i}} + (1 - \ell_i) n_{R_A}) \right)
\]

(86)

\[
\min \left( \max_{i \in \mathcal{U}} (\ell_i n_{A,R} + (1 - \ell_i) n_{B,R}), \right. 
\left. \max_{i \in \mathcal{U}} (\ell_i n_{R_{B_i}} + (1 - \ell_i) n_{R_A}) \right)
\]

(87)

where the first step is true since \( R \) satisfies the cut-set bound with \( \mathcal{U} = \mathcal{U} \cup \{1\} \) and \( \ell_1 = 1 \). Combining (68) and (86), we get \( R_{A_i} \leq 0 \), which is a contradiction to our assumption of \( R_{A_i} \geq 1 \). Similarly, if \( n_{R,B} \geq n_{R,B} \), we can write
where the first step is true since $R$ satisfies the cut-set bound \( (2) \) with $\mathcal{U} = \{1\}$ and $\ell_1 = 0$. Combining \( (68) \) and \( (69) \), we get $R_{B_1} \leq 0$, which is a contradiction to our assumption of $R_{A_1} \geq 1$.

4) $\exists j \in \{2, \ldots, M\}$ such that, $j \in \mathcal{U}$, $\ell_j = 0$, and
\[
\begin{align*}
n'_{RA_j} &= \max \left( \ell_i n'_{RB_R} + (1 - \ell_i) n'_{RA_i} \right), \\
n'_{RA_j} &= \min \left( \max \left( \ell_i n'_{RA_i} + (1 - \ell_i) n'_{RB_i} \right), \\
&\quad \quad \quad \max \left( \ell_i n'_{RB_R} + (1 - \ell_i) n'_{RA_i} \right) \right) \\
n'_{RA_j} &= n_{RA_j} - 1.
\end{align*}
\]

The proof that this case can not also happen is very similar to the previous case, hence we omit repetition.

**APPENDIX B**

In this Appendix we prove that the reduced rate-tuple $R' = (R_{A_1} - 1, 0, R_{A_2}, \ldots, R_{A_M}, 0)$, with $R_{A_1} \geq 1$, created in case 2 of the proof of Theorem 1 is in the cut-set region of the reduced network (defined in \( (11)-(14) \)). Since $R_{B_1} = \ldots = R_{B_M} = 0$, we just need to show that
\[
\sum \limits_{i \in \mathcal{U}} R'_{A_i} \leq \min \left( \max \left( n'_{A_i, R}, \max \limits_{i \in \mathcal{U}} n'_{RB_i} \right) \right), \\
\forall \mathcal{U} \subseteq \{1, \ldots, M\}.
\]

If $1 \in \mathcal{U}$, we have
\[
\sum \limits_{i \in \mathcal{U}} R'_{A_i} \overset{(1\in \mathcal{U})}{=} \sum \limits_{i \in \mathcal{U}} R_{A_i} - 1 \overset{(9)}{=} \min \left( \max \left( n_{A_i, R}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) - 1 = \min \left( \max \left( n_{A_i, R} - 1, \max \limits_{i \in \mathcal{U}} n_{RB_i} - 1 \right) \right) \overset{(11)}{=} \min \left( \max \left( n'_{A_i, R}, \max \limits_{i \in \mathcal{U}} n'_{RB_i} \right) \right).
\]

If $1 \notin \mathcal{U}$, we have
\[
\sum \limits_{i \in \mathcal{U}} R'_{A_i} \overset{(1\notin \mathcal{U})}{=} \sum \limits_{i \in \mathcal{U}} R_{A_i} \overset{(9)}{=} \min \left( \max \left( n_{A_i, R}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) = \min \left( \max \left( n_{A_i, R} - 1, \max \limits_{i \in \mathcal{U}} n_{RB_i} - 1 \right) + 1 \right) \overset{(11)}{=} \min \left( \max \left( n'_{A_i, R}, \max \limits_{i \in \mathcal{U}} n'_{RB_i} \right) + 1 \right).
\]

Therefore, if $1 \notin \mathcal{U}$, the only way to violate the cut-set bound is to have all above inequalities as equality, i.e.,
\[
\sum \limits_{i \in \mathcal{U}} R_{A_i} = \min \left( \max \left( n_{A_i, R}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) = \min \left( \max \left( n'_{A_i, R}, \max \limits_{i \in \mathcal{U}} n'_{RB_i} \right) + 1 \right).
\]

However, we show that this is in contradiction to our assumption of $R_{A_1} \geq 1$. To see this, note that by \( (11)-(14) \), the equality in \( (88) \) and \( (89) \) happens only if we have one of the following two cases.

1) $\exists j \in \{2, \ldots, M\}$ such that, $j \in \mathcal{U}$ and
\[
\begin{align*}
n'_{A_j, R} &= \max \left( n_{A_j, R}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right), \\
n'_{A_j, R} &= \min \left( \max \left( n'_{A_j, R}, \max \limits_{i \in \mathcal{U}} n'_{RB_i} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right), \\
n'_{A_j, R} &= \min \left( \max \left( n_{A_j, R}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right), \quad \min \left( \max \left( n_{A_j, R}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) \geq 1 \\
n'_{A_j, R} &= n_{A_j, R} - 1.
\end{align*}
\]

Since $n'_{A_j, R} = n_{A_j, R} - 1$, we have $n_{A_j, R} \geq l_u = n_{A_1, R}$.

Hence, we can write
\[
\begin{align*}
R_{A_1} + \sum \limits_{i \in \mathcal{U}} R_{A_i} &\leq \min \left( \max \left( n_{A_1, R}, \max \limits_{i \in \mathcal{U}} n_{A_i, R} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) \\
&\min \left( \max \left( n_{A_1, R}, \max \limits_{i \in \mathcal{U}} n_{A_i, R} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) \\
&\max \left( \max \left( n_{A_1, R}, \max \limits_{i \in \mathcal{U}} n_{A_i, R} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) \\
&\max \left( \max \left( n_{A_1, R}, \max \limits_{i \in \mathcal{U}} n_{A_i, R} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) \geq 1 \\
&\max \left( \max \left( n_{A_1, R}, \max \limits_{i \in \mathcal{U}} n_{A_i, R} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) \geq 1 \\
&\max \left( \max \left( n_{A_1, R}, \max \limits_{i \in \mathcal{U}} n_{A_i, R} \right), \max \left( n_{RB_i}, \max \limits_{i \in \mathcal{U}} n_{RB_i} \right) \right) \geq 1.
\end{align*}
\]

The proof that this case can not also happen is very similar to the previous case, hence we omit repetition.

**APPENDIX C**

**PROOF OF LEMMA 2**

Since the proof for both pairs are similar, we only bring the proof for pair $i = 1$. We claim that if $|h_{B_i, R}| > |h_{A_i, R}|$ and $R \in \mathcal{E}_u$, then $R \in \bar{\mathcal{E}}_u$, where $\bar{\mathcal{E}}_u$ is the up-link cut-set region of the network resulted by weakening $|h_{B_i, R}|$ and setting it equal to $|h_{A_i, R}|$. We call the new (undermined) uplink channel gains $(\bar{h}_{A_i, R}, h_{B_i, R}, \bar{h}_{A_i, R}, \bar{h}_{B_i, R})$. The claim is justified by check marking equations \( (26) \) to \( (31) \) for new capacities (with infinite down-link channel gains). The only non-obvious
inequalities are the ones in which $\hat{h}_{B_1}$ appears. By symmetry we only have to verify that (22) and (31) hold. Start with the original equations for $\tilde{h}_{A_1,R}$, $\tilde{h}_{A_2,R}$, $h_{A_2,R}$, and $h_{B_1,R}$ and note that the LHS of equations (22) and (31) are less than or equal to the LHS of (28) and (29) respectively and thus less than their RHS. Now replace $h_{A_1,R}$ with $\hat{h}_{B_1}$ and $h_{A_2,R}$ with $\hat{h}_{A_2}$ to get the desired inequalities. A similar argument on the downlink cut-set region shows that we can make the downlink channel gains of each pair consistent (in ordering) with the transmission rate and this completes the proof.

APPENDIX D

Decoding at the Relay

We receive the following signal at the relay

$$y_R = h_{A_1,R} \sqrt{\alpha^{(1)}_{A_1} x_{A_1}^1} + h_{A_2,R} \sqrt{\alpha^{(2)}_{A_2} x_{A_2}^1} + h_{B_1,R} x_{B_1}^1 + h_{A_2,R} \sqrt{\alpha^{(2)}_{A_2} x_{A_2}^2} + h_{B_1,R} x_{B_2}^1 + z_R.$$  

For the case considered here ($|h_{A_1,R}| \geq |h_{B_1,R}| \geq |h_{A_2,R}| \geq |h_{B_2,R}|$), we have the following decoding order at the relay: $x_{A_1}^1 \rightarrow t \rightarrow x_{A_2}^1 \rightarrow f$. It follows the decoding of the signals from pair $(A_1, B_1)$. Decoding of $x_{A_1}^1$ can be done with low error probability as long as

$$R^{(1)}_{A_1} \leq C \left( \frac{|h_{A_1,R}|^2 P \alpha^{(1)}_{A_1}}{2 \alpha^{(2)}_{B_1} |h_{B_1,R}|^2 P + \alpha^{(2)}_{A_2} |h_{A_2,R}|^2 P + 2 \alpha^{(2)}_{B_2} |h_{B_2,R}|^2 P + 1} \right).$$

Once $x_{A_1}^1$ is decoded, it can be subtracted successfully from the received signal. Thus, we have

$$\hat{y}_R = h_{B_1,R} \sqrt{\alpha^{(2)}_{B_1} (x_{A_1}^2 + x_{B_1}^2)} + h_{A_2,R} \sqrt{\alpha^{(1)}_{A_2} x_{A_2}^2} + h_{B_2,R} \sqrt{2 \alpha^{(2)}_{B_2} (x_{A_2}^2 + x_{B_2}^2)} + z_R.$$  

Next, the sum codeword $t$ of the lattice codes from $x_{A_1}^2$ and $x_{B_1}^1$ is decoded. The decoding of $t$ can be done with low error probability as long as

$$R^{(2)}_{A_1}, R_{B_1} \leq \log \left( \frac{|h_{B_1,R}|^2 P \alpha^{(2)}_{B_1}}{\alpha^{(1)}_{A_2} |h_{A_2,R}|^2 P + 2 \alpha^{(2)}_{A_2} |h_{B_2,R}|^2 P + 1} \right)^{+}.$$  

Once $t$ is decoded, it can be subtracted successfully from the received signal. Thus, we have

$$\hat{y}_R = h_{A_2,R} \sqrt{\alpha^{(1)}_{A_2} x_{A_2}^1} + h_{B_2,R} \sqrt{\alpha^{(2)}_{B_2} f} + z.$$  

It follows the decoding of the signals from pair $(A_2, B_2)$, beginning with the decoding of the Gaussian $x_{A_2}^1$. This can be done with low probability as long as

$$R^{(1)}_{A_2} \leq C \left( \frac{|h_{A_2,R}|^2 P \alpha^{(1)}_{A_2}}{2 |h_{B_2,R}|^2 P \alpha^{(2)}_{B_2} + 1} \right).$$

Once $x_{A_2}^1$ is decoded, it can be subtracted successfully from the received signal. Thus, we have

$$\hat{y}_R = h_{B_2,R} \sqrt{\alpha^{(2)}_{B_2} f} + z.$$  

As a final step, we want to decode the lattice point $f$. This can be done with low probability as long as

$$R_{B_2} \leq \left( \log \left( \frac{\alpha^{(2)}_{B_2} |h_{B_2,R}|^2 P}{\alpha_{B_2}^2} \right) \right)^{+}.$$  

APPENDIX E

Proof of Lemma 3

The three cases we have to consider are given in sections IV-C1 to IV-C3. In the following we provide the proof for each case separately.

A. Case $|h_{A_1,R}| \geq |h_{B_1,R}| \geq |h_{A_2,R}| \geq |h_{B_2,R}|$

Consider a 4-tuple $(r_{A_1}, r_{B_1}, r_{A_2}, r_{B_2})$ satisfying (47)–(52). Starting with (36), we equate

$$\left( \log \left( \frac{\alpha^{(2)}_{B_2} |h_{B_2,R}|^2 P}{\alpha_{B_2}^2} \right) \right)^{+} = r_{B_2} \Rightarrow \alpha_{B_2} = \frac{2^r_{B_2}}{|h_{B_2,R}|^2 P}. \tag{96}$$

Now from (48) we know that

$$\alpha_{B_2} \leq \frac{1 + |h_{B_2,R}|^2 P |h_{B_2,R}|^2 P + 1}{2 |h_{B_2,R}|^2 P} \leq 1,$$

which shows that this is a valid choice of $\alpha_{B_2}$. Next we equate $r_{A_2} - r_{B_2} = \text{RHS of (40)}$ and use (96). We get

$$\alpha_{A_2} = \frac{2^{r_{A_2} - r_{B_2} - 1}}{|h_{A_2,R}|^2 P} \tag{97}.$$

Using (33) and adding this to (97) we get

$$\alpha_{A_2}^{(1)} + \alpha_{A_2}^{(2)} = \frac{2^{r_{A_2} + r_{B_2} - 2} - 2^{r_{B_2} - 2}}{|h_{A_2,R}|^2 P} \leq \frac{3 \cdot 2^{r_{A_2}} - 2}{|h_{A_2,R}|^2 P} \leq 1,$$

verifying that this is a valid choice of $\alpha_{A_2}^{(1)}$, $\alpha_{A_2}^{(2)}$. Then we equate $r_{B_1} = \text{RHS of (45)}$, by setting

$$\alpha_{B_1}^{(2)} = \frac{2^{r_{B_1}} - 2^{r_{B_2} - 1} - 2^{r_{B_2} - 2}}{|h_{B_1,R}|^2 P} \tag{98}.$$

Using (33) and (98) and adding this to (99) we get

$$\alpha_{A_1}^{(1)} + \alpha_{A_1}^{(2)} \leq \frac{5 \cdot 2^{r_{A_1} + r_{B_2} - 1} + 2^{r_{A_1} + r_{B_2} - 2} - 3}{|h_{A_1,R}|^2 P} \leq 1,$$

which shows that this is a valid choice of $\alpha_{A_1}^{(1)}$, $\alpha_{A_1}^{(2)}$. 

B. Case $|h_{A,R}| \geq |h_{A_2,R}| \geq |h_{B,R}| \geq |h_{B_2,R}|$

Consider a 4-tuple $(r_{A_1}, r_{B_1}, r_{A_2}, r_{B_2})$ satisfying (47)–(52). Starting with (41), we equate

$$\left( \log \left( \frac{\alpha_B^2}{h_{B_2,R}^2 P} \right) \right)^+ = r_{B_2} \Rightarrow \alpha_B^2 = \frac{2^r_{B_2}}{|h_{B_2,R}|^2 P},$$

(100)

Now from (48) we know that

$$\alpha_B^2 \leq \frac{1 + |h_{B_2,R}|^2 P |h_{B_2,R}|^2 P \geq 1}{2|h_{B_2,R}|^2 P} \leq 1,$$

which shows that this is a valid choice of $\alpha_B^2$. Next we equate $r_{B_1}$ = RHS of (49), by setting

$$\alpha_B^2 = \frac{2^r_{B_1} (2 \cdot 2^r_{B_2} + 1)}{|h_{B_2,R}|^2 P} \leq 1,$$

(101)

verifying that this is a valid choice of $\alpha_B^2$. Then we equate $r_{B_2} - r_{B_2} = RHS$ of (33), by setting

$$\alpha_{A_1} = \frac{(2^r_{A_1} - r_{A_1} - 1) (4 \cdot 2^r_{A_2} + r_{A_2} + 2 (2^r_{A_1} + 2^r_{A_2}) + 1)}{|h_{A_1,R}|^2 P}$$

(102)

Using (33), $2^r x + 2^y \leq 2^{x+y}$ with $x, y \geq 1$, and (100) and adding this to (102) we get

$$\alpha_{A_1} + \alpha_{A_2} \leq \frac{6 \cdot 2^r_{A_2} + r_{A_2} + 2 (2^r_{A_1} + 2^r_{A_2}) + 1)}{|h_{A_1,R}|^2 P} \leq 1,$$

(103)

verifying that this is a valid choice of $\alpha_{A_1}, \alpha_{A_2}$. Now we equate $r_{A_1} - r_{B_1}$ = RHS of (37), by setting

$$\alpha_{A_1} = \frac{(2^r_{A_1} - r_{A_1} - 1) (4 \cdot 2^r_{A_2} + r_{A_2} + 2 (2^r_{A_1} + 2^r_{A_2}) + 1)}{|h_{A_1,R}|^2 P}$$

(104)

Using (33), $2^r x + 2^y \leq 2^{x+y}$ with $x, y \geq 1$, and (101) and adding this to (103) we get

$$\alpha_{A_1} + \alpha_{A_2} \leq \frac{2^r_{A_2} (2^r_{A_1} + 1)}{|h_{A_1,R}|^2 P} \leq \frac{|h_{A_1,R}|^2 P}{2|h_{A_1,R}|^2 P} \leq 1,$$

(106)

verifying that this is a valid choice of $\alpha_{A_1}, \alpha_{A_2}$. Then we equate $r_{A_2} - r_{B_2}$ = RHS of (33), by setting

$$\alpha_{A_2} = \frac{(2^r_{A_1} - r_{A_1} - 1) (4 \cdot 2^r_{A_2} + r_{A_2} + 2 (2^r_{A_1} + 2^r_{A_2}) + 1)}{|h_{A_1,R}|^2 P}$$

(107)

Using (33), $2^r x + 2^y \leq 2^{x+y}$ with $x, y \geq 1$, and (106) and adding this to (107) we get

$$\alpha_{A_1} + \alpha_{A_2} \leq \frac{2^r_{A_2} (2^r_{A_1} + 1)}{|h_{A_1,R}|^2 P} \leq \frac{|h_{A_1,R}|^2 P}{2|h_{A_1,R}|^2 P} \leq 1,$$

(108)

verifying that this is a valid choice of $\alpha_{A_1}, \alpha_{A_2}$. Now we equate $r_{A_1} - r_{B_1}$ = RHS of (32), by setting

$$\alpha_{A_1} = \frac{(2^r_{A_1} - r_{A_1} - 1) (4 \cdot 2^r_{A_2} + r_{A_2} + 2 (2^r_{A_1} + 2^r_{A_2}) + 1)}{|h_{A_1,R}|^2 P}.$$
Using $2^x + 2^y \leq 2^{x+y}$ with $x, y \geq 1$, and (105) and adding this to (108) we get

$$
\alpha_{A_1}^{(1)} + \alpha_{A_1}^{(2)} \leq (2^{R_{A_1} - r_{B_1}} - 1) \times \frac{4 \cdot 2^{r_{B_2} r_{B_1} + 2 (2^{r_{B_1}} + 2^{r_{B_2}}) + 1}}{\|h_{A_1 R}|^2 P}
$$

Finally we equate $r_{A_1} - r_{B_1} = r_{A_2} - r_{B_2}$ as noise. This can be done successfully with low probability of error as long as

$$
R_{R_1}^{(2)} \leq C \left( \frac{|h_{R B_1}|^2 P \alpha_R^{(2)}}{1 + |h_{R B_1}|^2 P \alpha_R^{(1)}} \right)
$$

which shows that this is a valid choice of $\alpha_{A_1}^{(1)}, \alpha_{A_1}^{(2)}$.

Using $2^x + 2^y \leq 2^{x+y}$ with $x, y \geq 1$, and (106) and adding this to (109) we get

$$
\alpha_{A_1}^{(1)} + \alpha_{A_1}^{(2)} = (2^{r_{A_1} - r_{B_1} r_{A_2} - r_{B_2}} - 2^{r_{A_2} - r_{B_2}}) \times \frac{4 \cdot 2^{r_{B_2} r_{B_1} + 2 (2^{r_{B_1}} + 2^{r_{B_2}}) + 1}}{\|h_{A_1 R}|^2 P}
$$

Finally, $x^{(1)}$ is decoded by treating $x_R^{(1)}$ as noise. This can be done successfully with low probability of error as long as

$$
R_{R_1}^{(2)} \leq C \left( \frac{|h_{R B_1}|^2 P \alpha_R^{(2)}}{1 + |h_{R B_1}|^2 P \alpha_R^{(1)}} \right)
$$

which shows that this is a valid choice of $\alpha_{A_1}^{(1)}, \alpha_{A_1}^{(2)}$.

**APPENDIX F**

**DECODING AT THE NODES**

With $R_{R_1}^{(1)} = R_{R_1}^{(1)}$, $R_{R_1}^{(2)} = R_{R_1}^{(2)}$, $R_{R_1}^{(3)} = R_{R_1}^{(3)}$, $R_{R_1}^{(4)} = R_{R_1}^{(4)}$, we describe the decoding strategies at the nodes and the achievable rates for the case $|h_{R B_1}| \geq |h_{R A_1}| \geq |h_{R B_2}| \geq |h_{R A_2}|$.

**A. Decoding at node $B_1$**

The node $B_1$ first decodes $x^{(4)}$ (corresponds to $f$ from the uplink) by treating $x_R^{(4)}$ as noise. This can be done with low probability of error as long as

$$
R_{R_1}^{(4)} \leq C \left( \frac{|h_{R B_1}|^2 P \alpha_R^{(4)}}{1 + |h_{R B_1}|^2 P \sum_{j=1}^3 \alpha_R^{(j)}} \right)
$$

Once decoded, the signal $x^{(4)}$ is canceled from the received signal and $x^{(3)}$ is decoded by treating $x^{(3)}$ as noise. This can be done successfully with low probability of error as long as

$$
R_{R_1}^{(3)} \leq C \left( \frac{|h_{R B_1}|^2 P \alpha_R^{(3)}}{1 + |h_{R B_1}|^2 P \sum_{j=1}^2 \alpha_R^{(j)}} \right)
$$

Once decoded, the signal $x^{(3)}$ is canceled from the received signal and $x^{(2)}$ (corresponds to $t$ from the uplink) is decoded by treating $x^{(2)}$ as noise. This can be done successfully with low probability of error as long as

$$
R_{R_1}^{(2)} \leq C \left( \frac{|h_{R B_1}|^2 P \alpha_R^{(2)}}{1 + |h_{R B_1}|^2 P \sum_{j=1}^2 \alpha_R^{(j)}} \right)
$$

**B. Decoding at node $A_1$**

The node $A_1$ proceeds similarly with the exception that $x_R^{(1)}$ is known already and can be canceled from the received signal. After having decoded $x_R^{(3)}$ and $x_R^{(4)}$, $x_R^{(2)}$ is decoded free of interference. This can be done with low probability of error as long as

$$
R_{R_1}^{(2)} \leq C \left( \frac{|h_{R A_1}|^2 P \alpha_R^{(2)}}{1 + |h_{R A_1}|^2 P \sum_{j=1}^2 \alpha_R^{(j)}} \right)
$$

**C. Decoding at node $B_2$**

The receivers of the second pair have the same order of detection. Thus, the node $B_2$ can decode $R_{R_1}^{(4)}$ with low probability of error as long as

$$
R_{R_1}^{(4)} \leq C \left( \frac{|h_{R B_2}|^2 P \alpha_R^{(4)}}{1 + |h_{R B_2}|^2 P \sum_{j=1}^3 \alpha_R^{(j)}} \right)
$$

Once decoded, the signal $x_R^{(4)}$ is canceled from the received signal and $x_R^{(3)}$ is decoded by treating $x_R^{(1)}$ and $x_R^{(2)}$ as noise. This can be done successfully with low probability of error as long as

$$
R_{R_1}^{(3)} \leq C \left( \frac{|h_{R B_2}|^2 P \alpha_R^{(3)}}{1 + |h_{R B_2}|^2 P \sum_{j=1}^2 \alpha_R^{(j)}} \right)
$$

**D. Decoding at node $A_2$**

Assuming that the node $A_2$ knows the strategy of the relay and the codebook it has used, it can reconstruct $x^{(3)}$ perfectly, since it contains only its own message. Thus, it cancels the effect of $x_R^{(3)}$ from the received signal. As a next and final step, it decodes $x^{(4)}$. This can be done with low probability of error as long as

$$
R_{R_1}^{(4)} \leq C \left( \frac{|h_{R A_2}|^2 P \alpha_R^{(4)}}{1 + |h_{R A_2}|^2 P \sum_{j=1}^2 \alpha_R^{(j)}} \right)
$$

Thus, in summary we have

$$
R_{R_1}^{(4)} \leq \min (\text{RHS of } (112), \text{RHS of } (113))
$$

and

$$
R_{R_1}^{(2)} \leq \min (\text{RHS of } (111), \text{RHS of } (110))
$$
APPENDIX G

PROOF OF LEMMA 4

The three cases we have to consider are given in sections IV-E1-IV-E3. In the following we provide the proof for each case separately.

A. Case $|h_{RB_1}| \geq |h_{RA_2}| \geq |h_{RB_2}| \geq |h_{RA_1}|$

Consider a 4-tuple $(r_{A_1}, r_{B_1}, r_{A_2}, r_{B_2})$ satisfying (62)-(67). Starting with the first equation in (55), we equate

$$\log \left( 1 + \alpha_R^{(1)} |h_{RB_1}|^2 P \right) = r_{A_1} - r_{B_1}$$

$$\Rightarrow \alpha_R^{(1)} = \frac{2^{r_{A_1} - r_{B_1}} - 1}{|h_{RB_1}|^2 P}.$$  

Now from (62) we know that

$$\alpha_R^{(1)} \leq \frac{1 + |h_{RB_2}|^2 P}{4 |h_{RB_1}|^2 P} - 1 \leq 1,$$

which shows that this is a valid choice of $\alpha_R^{(1)}$.

From (53), we have

$$R_{A_1}, R_{B_1} \leq C \left( \frac{|h_{RB_1}|^2 P \alpha_R^{(2)}}{1 + |h_{RB_1}|^2 P \alpha_R^{(1)}} \right).$$  

Next we equate $r_{B_1}$ = RHS of (115), by setting

$$\alpha_R^{(2)} = \frac{(2^{r_{B_1}} - 1)}{|h_{RB_1}|^2 P}.$$  

Using (62) and (114) and adding this to (116) we get

$$\alpha_R^{(1)} + \alpha_R^{(2)} \leq \frac{1 + |h_{RB_2}|^2 P}{4 |h_{RB_1}|^2 P} - 1 \leq 1,$$

verifying that this is a valid choice of $\alpha_R^{(1)}, \alpha_R^{(2)}$. Then we equate $r_{A_2} - r_{B_2}$ = RHS of (55) (second equation), by setting

$$\alpha_R^{(3)} = \frac{(2^{r_{A_2} - r_{B_2}} - 1)}{|h_{RB_2}|^2 P}.$$  

Using (63), (65) and (117) and adding this to (118) we get

$$\sum_{j=1}^{3} \alpha_R^{(j)} \leq \frac{1 + |h_{RB_2}|^2 P}{4 |h_{RB_1}|^2 P} - 1 + \frac{1 + |h_{RB_1}|^2 P}{4 |h_{RB_1}|^2 P} - 1 \leq 1,$$

verifying that this is a valid choice of $\alpha_R^{(j)}, j = 1 \ldots 3$. Finally from (54), we have

$$R_{A_2}, R_{B_2} \leq C \left( \frac{|h_{RA_1}|^2 P \alpha_R^{(4)}}{1 + |h_{RA_1}|^2 P (\alpha_R^{(1)} + \alpha_R^{(2)})} \right).$$  

Thus, we equate $r_{B_2}$ = RHS of (119), by setting

$$\alpha_R^{(4)} = \frac{(2^{r_{B_2}} - 1)}{|h_{RA_1}|^2 P} \left( 1 + \frac{|h_{RA_1}|^2 P}{|h_{RB_2}|^2 P} (2^{r_{A_1}} - 1) \right).$$  

Using (63), (65), (114), (116), (118) and adding this to (120) we get

$$\sum_{j=1}^{4} \alpha_R^{(j)} \leq \frac{1 + |h_{RA_2}|^2 P}{4 |h_{RA_2}|^2 P} - 1 + \frac{1 + |h_{RB_2}|^2 P}{4 |h_{RB_2}|^2 P} - 1 \leq 1,$$

which shows that this is a valid choice of $\alpha_R^{(j)}, j = 1 \ldots 4$.

B. Case $|h_{RB_1}| \geq |h_{RA_2}| \geq |h_{RB_2}| \geq |h_{RA_1}|$

Consider a 4-tuple $(r_{A_1}, r_{B_1}, r_{A_2}, r_{B_2})$ satisfying (62)-(67). Starting with the first equation in (55), we equate

$$\log \left( 1 + \alpha_R^{(1)} |h_{RB_1}|^2 P \right) = r_{A_1} - r_{B_1}$$

$$\Rightarrow \alpha_R^{(1)} = \frac{2^{r_{A_1} - r_{B_1}} - 1}{|h_{RB_1}|^2 P}.$$  

Now from (62) we know that

$$\alpha_R^{(1)} \leq \frac{1 + |h_{RB_2}|^2 P}{4 |h_{RB_1}|^2 P} - 1 \leq 1,$$

which shows that this is a valid choice of $\alpha_R^{(1)}$.

Next we equate $r_{A_2} - r_{B_2}$ = RHS of (55) (second equation), by setting

$$\alpha_R^{(3)} = \frac{(2^{r_{A_2} - r_{B_2}} - 1)}{|h_{RB_2}|^2 P} \left( 1 + \frac{|h_{RA_2}|^2 P}{|h_{RB_2}|^2 P} (2^{r_{A_1}} - 1) \right).$$  

Using (63), (65) and (117) and adding this to (122) we get

$$\alpha_R^{(1)} + \alpha_R^{(3)} \leq \frac{1 + |h_{RB_2}|^2 P}{4 |h_{RB_1}|^2 P} - 1 + \frac{1 + |h_{RB_1}|^2 P}{4 |h_{RB_1}|^2 P} - 1 \leq 1,$$

verifying that this is a valid choice of $\alpha_R^{(1)}, \alpha_R^{(3)}$.

From (56), we have

$$R_{A_1}, R_{B_1} \leq C \left( \frac{|h_{RA_1}|^2 P \alpha_R^{(2)}}{1 + |h_{RA_1}|^2 P (\alpha_R^{(1)} + \alpha_R^{(3)})} \right).$$  

Next we equate $r_{B_1}$ = RHS of (123), by setting

$$\alpha_R^{(2)} = \frac{(2^{r_{B_1}} - 1)}{|h_{RA_1}|^2 P}.$$  

Using (63) and (121) and adding this to (124) we get

$$\sum_{j=1}^{3} \alpha_R^{(j)} \leq \frac{2^{r_{B_1}} - 1}{|h_{RA_1}|^2 P} + \frac{2^{r_{A_1} + r_{B_2}} - 1}{|h_{RB_2}|^2 P} + \frac{2^{r_{A_1} + r_{B_2}} - 1}{|h_{RB_1}|^2 P} \leq 1,$$

verifying that this is a valid choice of $\alpha_R^{(j)}, j = 1 \ldots 3$. Finally from (54), we have

$$R_{A_2}, R_{B_2} \leq C \left( \frac{|h_{RA_1}|^2 P \alpha_R^{(4)}}{1 + |h_{RA_1}|^2 P (\alpha_R^{(3)} + \alpha_R^{(2)})} \right).$$  


Thus, we equate \( r_{B_2} \) RHS of (126), by setting
\[
\alpha_{R}^{(4)} = \frac{(2^{r_{B_2}} - 1) (2^{r_{A_1}} + (2^{r_{A_1}} - 1) |h_{RA_1}|^2 P \alpha_{R}^{(3)})}{|h_{RA_1}|^2 P}
\]
(127)
Using (63), (65), (121), (124), (122) and adding this to (127) we get
\[
\sum_{j=1}^{4} \alpha_{R}^{(j)} \leq \frac{1 + |h_{RA_1}|^2 P}{|h_{RB_1}|^2 P} + \frac{1 + |h_{RB_1}|^2 P - 1}{|h_{RB_1}|^2 P} - 1
\]
+ \frac{1 + |h_{RB_2}|^2 P}{|h_{RB_2}|^2 P} - 1 \leq 1
\]
which shows that this is a valid choice of \( \alpha_{R}^{(j)} \), \( j = 1 \ldots 4 \).

C. Case \( |h_{RB_1}| \geq |h_{RB_2}| \geq |h_{RA_1}| \)

Consider a 4-tuple \((r_{A_1}, r_{B_1}, r_{A_2}, r_{B_2})\) satisfying (63)-(67).
Starting with the first equation in (58), we equate
\[
\log \left( 1 + \alpha_{R}^{(1)} |h_{RB_1}|^2 P \right) = r_{A_1} - r_{B_1}
\]
(128)
\[
\Rightarrow \alpha_{R}^{(1)} = \frac{2^{r_{A_1}} - r_{B_1} - 1}{|h_{RB_1}|^2 P}
\]
Now from (62) we know that
\[
\alpha_{R}^{(1)} \leq \frac{1 + |h_{RB_1}|^2 P}{|h_{RB_1}|^2 P} - 1 \leq 1,
\]
which shows that this is a valid choice of \( \alpha_{R}^{(1)} \).
Next we equate \( r_{B_2} - r_{B_2} = \) RHS of (61) (second equation), by setting
\[
\alpha_{R}^{(3)} = \frac{(2^{r_{A_2}} - r_{B_2} - 1) \left( 1 + \frac{|h_{RB_2}|^2 P}{|h_{RB_2}|^2 P} + (2^{r_{A_1}} - r_{B_1} - 1) \right)}{|h_{RB_2}|^2 P}
\]
(129)
Using (63), (64) and (17) and adding this to (129) we get
\[
\alpha_{R}^{(1)} + \alpha_{R}^{(3)} \leq \frac{1 + |h_{RB_1}|^2 P}{|h_{RB_2}|^2 P} - 1 + \frac{1 + |h_{RB_1}|^2 P}{|h_{RB_2}|^2 P} - 1 \leq 1,
\]
which verifies this is a valid choice of \( \alpha_{R}^{(1)}, \alpha_{R}^{(3)} \).
From (60), we have
\[
R_{A_2}^{(2)}, R_{B_2} \leq C \left( \frac{|h_{RA_2}|^2 P \alpha_{R}^{(4)}}{1 + |h_{RA_2}|^2 P \alpha_{R}^{(1)}} \right)
\]
(130)
Next we equate \( r_{B_2} \) RHS of (130), by setting
\[
\alpha_{R}^{(4)} = \frac{(2^{r_{B_2}} - 1) \left( 1 + \frac{|h_{RA_2}|^2 P \alpha_{R}^{(3)}}{|h_{RA_2}|^2 P} \right)}{|h_{RA_2}|^2 P}
\]
(131)
Using (62) and (128) and adding this to (131) we get
\[
\alpha_{R}^{(1)} + \alpha_{R}^{(3)} + \alpha_{R}^{(4)} \leq \frac{2^{r_{B_2}} - 1 + \frac{2^{r_{A_2}} - r_{B_2} - 1}{|h_{RB_2}|^2 P} + \frac{2^{r_{A_1}} - r_{B_1} - 1}{|h_{RB_1}|^2 P}}{|h_{RB_2}|^2 P} + \frac{2^{r_{A_1}} - r_{B_2} - 1}{|h_{RB_1}|^2 P} - 1 \leq 1,
\]
verifying that this is a valid choice of \( \alpha_{R}^{(1)}, \alpha_{R}^{(3)}, \alpha_{R}^{(4)} \).
Finally from (54), we have
\[
R_{A_1}^{(2)}, R_{B_1} \leq C \left( \frac{|h_{RA_1}|^2 P \alpha_{R}^{(2)}}{1 + |h_{RA_1}|^2 P (\alpha_{R}^{(3)} + \alpha_{R}^{(4)})} \right)
\]
(133)
Thus, we equate \( r_{B_1} = \) RHS of (133), by setting
\[
\alpha_{R}^{(2)} = \frac{(2^{r_{B_1}} - 1)}{|h_{RA_1}|^2 P} \left( 1 + \alpha_{R}^{(3)} + \alpha_{R}^{(4)} \right) |h_{RA_1}|^2 P
\]
(134)
Using (63), (65), (128), (131), (129) and adding this to (134) we get
\[
\sum_{j=1}^{4} \alpha_{R}^{(j)} \leq \frac{2^{r_{B_1}} - 1 + \frac{2^{r_{A_2}} + r_{B_2} - 1}{|h_{RB_2}|^2 P}}{|h_{RB_2}|^2 P} + \frac{2^{r_{A_1}} + r_{B_2} - 1}{|h_{RB_1}|^2 P} - 1 \leq 1,
\]
which shows that this is a valid choice of \( \alpha_{R}^{(j)} \), \( j = 1, \ldots, 4 \).

REFERENCES
[1] A.S. Astérmie, M.A. Kajhejnedaz, A. Sezgin, and B. Hassibi, “Capacity region of the deterministic multi-pair directional relay network,” IEEE ITW 2009, Volos, Greece, June 10-12, 2009.
[2] A. Sezgin, M.A. Kajhejnedaz, A.S. Astérmie, and B. Hassibi, “Approximate capacity region of the two-pair bi-directional gaussian relay network,” IEEE ISIT 2009, Seoul, Korea, June 28- July 3, 2009.
[3] A.S. Astérmie, S. Diggavi, and D. Tse, “Wireless network information flows: A deterministic approach,” IEEE Trans. on Info. Theory, vol. 57, no. 4, pp. 1872 – 1905, Apr. 2011.
[4] E. C. van der Meulen, “Three-terminal communication channels,” Ad. Appl. Pmb., vol. 3, pp. 120–154, September 1971.
[5] T.M. Cover and A.E. Gamal, “Capacity theorems for the relay channel,” IEEE Trans. on Info. Theory, vol. 25, no. 5, pp. 572–584, Sep. 1979.
[6] C. E. Shannon, “Two-way communication channels,” Proc. 4th Berkeley Symp. Math. Statistics Probability, Berkeley, CA, pp. 611–644, 1961.
[7] B. Rankov and A. Wittneben, “Achievable rate regions for the two-way relay channel,” ISIT, Seattle, USA, July 9-14, 2006.
[8] S.H. Lim, Y.-H. Kim, A. El Gamal, and S.-Y. Chung, “Noisy network coding,” submitted to IEEE Transactions on Information Theory, Available: http://arxiv.org/abs/1002.3188. Mar. 2010.
[9] K. Katti, H. Rahul, W. Hu, D. Katabi, M. Medard, and J. Crowcroft, “XORs in the Air: Iterative Wireless Network Coding,” ACM SIGCOMM, Pisa, Italy, September, 11-15 2006.
[10] C. Hausl and J. Hagenerau, “Iterative network and channel decoding for the two-way relay channel,” Proc. of IEEE ICC 2006, Istanbul, Turkey, June 2006.
[11] J-I. Baik and S.-Y. Chung, “Iterative network coding for two-way relay channels using lattices,” Proc. of IEEE ICC 2008, Beijing, China, pp. 3898–3902, May 2007.
[12] K. Narayanan, M. P. Wilson, and A. Sprintson, “Joint physical layer coding and network coding for bi-directional relaying,” Proc. of Allerton Conference on Communication, Control and Computing, 2007.
[13] W. Nam, S.-Y. Chung, and Y.H. Lee, “Capacity of the gaussian two-way relay channel to within 1/2 bit,” IEEE Trans. on Info. Theory, vol. 56, no. 11, pp. 5488 – 5494, July 2010.
[14] N. Lee, J.-B. Lim, and J. Chun, “Degrees of freedom of the MIMO Y channel: Signal space alignment for network coding,” IEEE Trans. on Info. Theory, vol. 56, no. 7, pp. 3332 – 3342, July 2010.
[15] B. Rankov and A. Wittneben, “Spectral efficient protocols for half-duplex fading relay channels,” IEEE Journal on Selected Areas in Communications, vol. 25, no. 2, pp. 379–389, February 2007.
[16] M. Chen and A. Yener, “Multisuer two-way relaying for interference limited sytems,” Proc. of IEEE ICC 2008, Beijing, China, May 2008.
[17] M. Chen and A. Yener, “Interference management for multisuer two-way relaying,” Proc. of IEEE CESS 2008, Princeton, NJ, March 2008.
[18] A. S. Astérmie, A. Sezgin, and D. Tse, “Approximate capacity of the two-way relay channel: A deterministic approach,” 46th Allerton Conf. On Comm., Control, and Computing, 2008.
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