Diversity, Coding, and Multiplexing Trade-Off of Network-Coded Cooperative Wireless Networks

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Abstract—In this paper, we study the performance of network-coded cooperative diversity systems with practical communication constraints. More specifically, we investigate the interplay between diversity, coding, and multiplexing gain when the relay nodes do not act as dedicated repeaters, which only forward data packets transmitted by the sources, but they attempt to pursue their own interest by forwarding packets which contain a network-coded version of received and their own data. We provide a very accurate analysis of the Average Bit Error Probability (ABEP) for two network topologies with three and four nodes, when practical communication constraints, i.e., erroneous decoding at the relays and fading over all the wireless links, are taken into account. Furthermore, diversity and coding gain are studied, and advantages and disadvantages of cooperation and binary Network Coding (NC) are highlighted. Our results show that the throughput increase introduced by NC is offset by a loss of diversity and coding gain. It is shown that there is neither a coding nor a diversity gain for the source node when the relays forward a network-coded version of received and their own data. Compared to other results available in the literature, the conclusion is that binary NC seems to be more useful when the relay nodes act only on behalf of the source nodes, and do not mix their own packets to the received ones. Analytical derivation and findings are substantiated through extensive Monte Carlo simulations.

I. INTRODUCTION

Cooperative/multi-hop networking has recently emerged as a strong candidate technology for many future wireless applications [1], [2]. The basic premise of cooperative/multi-hop communications is to achieve and to exploit the benefits of spatial diversity without requiring each mobile node to be equipped with co-located multiple antennas. On the contrary, each mobile node becomes part of a large distributed array and shares its single-antenna (as well as hardware, processing, and energy resources) to help other nodes of the network to achieve better performance/coverage. However, the efficient exploitation of cooperative/multi-hop networking is faced by the following challenges [3], [4]: i) due to practical considerations, such as the half-duplex constraint or to avoid interference caused by simultaneous transmissions, distributed cooperation needs extra bandwidth resources (e.g., time slots or frequencies), which might result in a loss of system throughput; ii) relay nodes are forced to use their own resources to forward the packets of other nodes, usually without receiving any rewards, except for the fact that the whole system can become more efficient; and iii) in classical cooperative protocols, the relay nodes that perform a retransmission on behalf of other nodes must delay their own frames, which has an impact on the latency of the network.

To overcome these limitations, a new technology named Network Coding (NC) has recently been introduced to improve the network performance [5]–[7]. NC can be broadly defined as an advanced routing or encoding mechanism at the network layer, which allows network nodes not only to forward but also to process incoming data packets. Different forms of NC exist in the literature, e.g., algebraic NC, physical-layer NC, and Multiple-Input-Multiple-Output (MIMO–) NC, which offer a different trade-off between achievable performance and implementation complexity. The interested reader might consult [4] for a recent survey and comparison of these methods. The common feature of all NC approaches is that the network throughput is improved by allowing some network nodes to combine many incoming packets, which, after being mixed, need a single wireless resource (e.g., a time slot or a frequency) for their transmission. Thus, NC is considered a potential and effective enabler to recover the throughput loss experienced by cooperative/multi-hop networking [3]. Theory and experiments have shown that network-coded cooperative/multi-hop systems can be extremely useful for wireless networks with disruptive channel and connectivity conditions [6], [7].

The performance of cooperative/multi-hop networks has been studied extensively during the last years, see, e.g., [8]–[11], and many important conclusions have been drawn about the achievable diversity and coding gain over fading channels. On the other hand, the analysis of the performance of cooperative/multi-hop systems with NC is almost unexplored so far. More specifically, understanding the interplay between the multiplexing gain introduced by NC and the achievable diversity/coding gain introduced by cooperation is an open and challenging research problem, especially when practical communication constraints (erroneous decoding and fading) are taken into account [12]–[14]. Some recent results on this matter are [15]–[21]. In particular, [16] and [20] have recently provided an accurate and closed-form analysis of network-coded cooperative/multi-hop systems by estimating both diversity and coding gain with realistic source–to–relay links. These papers have highlighted, for some network topologies and encoding schemes, the potential benefits of NC to recover the throughput loss of cooperative/multi-hop networking.

However, the analysis in [16] and [20] considers the classical scenario where some network nodes (i.e., the relays) operate only on behalf of other network nodes (i.e., the sources) when forwarding data to a given destination. In other words, the relays are dedicated network elements with no data to transmit and, thus, they receive no direct reward from
cooperation. In this paper, we are interested in studying the interplay between diversity, coding, and multiplexing gain of network-coded cooperative/multi-hop wireless networks when the relays have their own data packets to be transmitted to a common destination, and exploit NC to transmit them along with the packets that have to be relayed on behalf of the sources. This way, the relays can help the sources without the need to: i) delay the transmission of their own data packets; and ii) use specific resources (energy and processing) to forward the packets of the sources. Thus, NC can potentially avoid throughput and energy loss. However, it is not clear whether performing NC at the relay nodes entail any performance (i.e., diversity or coding gain) loss with respect to classical cooperative diversity. The main aim of this paper is to shed lights on this matter, and to highlight the fundamental diversity, coding, and multiplexing trade-off with realistic communication constraints and binary NC at the relays. To this end, two network topologies are considered with 3 (1 source, 1 relay, 1 destination) and 4 nodes (1 source, 2 relays, 1 destination), and the end–to–end Average Bit Error Probability (ABEP) over independent but non–identically distributed (i.n.i.d) Rayleigh fading channels is computed in closed–form. Our results highlight that the throughput increase introduced by NC is offset by a loss of the diversity gain. More specifically, it is shown that, when the relays forward a network–coded version of received and their own data packets, there is neither a coding nor a diversity gain for the source. Compared to other results available in the literature [16], [20], the conclusion is that binary NC seems to be more useful when the relays act on behalf of the sources only, and do not mix their own packets to the received ones.

The remainder of this paper is organized as follows. In Section II system model and problem statement are summarized. In Section III the analytical framework to compute the ABEP is described. In Section IV the achievable diversity, coding, and multiplexing gain of various schemes with and without NC are analyzed and compared. In Section V some numerical results are shown. Finally, Section VI concludes this paper.

II. SYSTEM MODEL AND PROBLEM STATEMENT

We study two cooperative network topologies with three and four nodes, as shown in Fig. 1 and Fig. 2 respectively. We consider a Time–Division–Multiple–Access (TDMA) protocol, where all transmissions take place in non–overlapping time–slots (Ts denotes the duration of a time–slot). Also, we assume the half–duplex constraint, i.e., nodes cannot transmit and receive at the same time [3]. Furthermore, we analyze the MIMO–NC approach, where network decoding and demodulation at the final destination are jointly performed at the physical layer, which results in a cross–layer decoding algorithm [4]. For analytical tractability, we assume that each node uses uncoded Binary Phase Shift Keying (BPSK) modulation. In those scenarios where NC is exploited, we consider binary NC (exclusive OR denoted by ⊕) as this provides a low–complexity design of the relays. Each wireless channel is assumed to experience Rayleigh fading. More specifically, the fading coefficient between two generic nodes X and Y is denoted by h_{XY}, and it is assumed to be a circular symmetric complex Gaussian Random Variable (RV) with zero mean and variance \sigma^2_{XY} per dimension. Fading over different links is assumed to be i.n.i.d to account for different propagation distances and shadowing effects. The noise at the input of node Y and related to the transmission from node X to node Y is denoted by n_{XY}, and it is assumed to be complex Additive White Gaussian (AWG) with variance \sigma^2_{0}/2 per dimension. Finally, n_{XY} at different time–slots or at the input of different nodes are assumed to be independent and identically distributed (i.i.d.).

A. Problem Statement

The main objective of this paper is to understand the performance vs. throughput trade–off provided by NC over fading channels. To be more specific, let us consider the 3–node scenario in Fig. 1. Similar comments apply to the 4–node scenario in Fig. 2. We have two nodes (S and R), which have data to transmit to node D. In Scenario (a), both nodes perform their transmission to D in a selfish mode, i.e., no cooperation. In Scenario (b), node R is willing to help node S to forward the overhead packet to node D. In this case, node S acts as a “golden user”, and node R delays the transmission of its own data packet to help node S first. In this case, node S can take advantage of cooperation to improve its performance. However, node R has to share its transmission energy with node S, and it must delay its own transmission: this is the price of cooperation. In Scenario (c), node R uses NC to avoid the limitations just mentioned. By using NC, node R can avoid to delay its own packet, and it can transmit a coded (XOR) version of overhead packet from node S and its own packet. The gain is twofold: i) no transmission delay; and ii)
no need to share transmission energy with node $S$. In this case, the overall transmission can be completed in two time-slots rather than in three time-slots as in Scenario (b). Thus the network throughput increases.

The fundamental questions we want to address in this paper are: i) Is there any performance (diversity/coding gain) loss, with respect to selfish and cooperative scenarios, for this throughput gain?; and ii) In case of performance loss, is this only due to erroneous decoding at node $R$ or is this related to NC operations too? Our closed-form asymptotic analysis will provide a clear answer to both questions. Similar questions hold for Fig. 3 as well, where we can see that, depending on the level of cooperation and NC, the throughput of the network, i.e., the number of time-slots, is different.

Due to space limitations, we are unable to provide a step-by-step analysis and derivation for all the scenarios shown in Fig. 1 and Fig. 2. However, the analytical development is very similar for all of them. Thus, for ease of exposition and clarity, we have decided to focus our attention on a scenario only. We have chosen Scenario (d) in Fig. 2 as it is the most general one. So, in the remainder of this paper only this scenario will be analyzed analytically. However, in Section IV we will summarize the final expression of the ABEP for all the scenarios in Fig. 1 and Fig. 2 and we will compare achievable performance and throughput of all of them.

**B. Signal Model**

Let us consider Scenario (d) in Fig. 2. During the first time-slot, node $S$ broadcasts a BPSK modulated bit, $x_S = \sqrt{E_m(1 - 2b_S)}$, where $E_m$ is the average transmitted energy and $b_S \in \{0, 1\}$ is the bit emitted by $S$. The signals received at nodes $R$, $T$, and $D$ are given by $y_{SX} = h_S x_S + n_{SX}$, where $x = R, T, X = X, T, X$, and $X = D$, respectively. Similar to [16], [19], [20], the intermediate nodes $R$ and $T$ demodulate the received bit by using conventional Maximum-Likelihood (ML-) optimum decoding:

$$
\hat{b}_S^{(X)} = \arg \min_{b_S \in \{0, 1\}} \left\{ |y_{SX} - \sqrt{E_m b_S x_S (1 - 2b_S)}|^2 \right\}
$$

where $X = R$ and $X = T$, and ($\cdot$) and ($\tilde{\cdot}$) denote detected/estimated and trial bit of the hypothesis-detection problem, respectively. $\hat{b}_S^{(X)}$ is the estimate of $b_S$ at node $X$.

During the second time-slot, node $R$ remodulates and forwards its estimate of $b_S$, i.e., $\hat{b}_S^{(R)}$, to node $D$. The transmitted bit is $x_R = \sqrt{E_m/2(1 - 2\hat{b}_S^{(R)})}$. Let us note that node $R$ uses only half of its available energy to forward $\hat{b}_S^{(R)}$ on behalf of node $S$, as it needs half energy to transmit its own data during the fourth time-slot. This allows us to consider a total energy constraint, and it guarantees a fair comparison among the scenarios. Similar considerations apply to all the scenarios shown in Fig. 1 and Fig. 2. The signal received at node $D$ is

$$
y_{RD} = h_{RD} x_R + n_{RD}.
$$

During the third time-slot, node $T$ performs similar operations as node $R$ in the second time-slot. However, node $T$ applies binary NC to avoid to use two time-slots to help nodes $S$ and to transmit its own data. More specifically, the bit transmitted by node $T$ is $x_T = \sqrt{E_m} \left( 1 - (\hat{b}_S^{(R)} \oplus b_T) \right)$, where $b_T$ is the bit that $T$ wants to transmit to node $D$. Unlike node $R$, node $T$ uses full transmission energy, since, with the help of NC, it does not need an extra time-slot to forward its own data.

The signal received at $D$ is $y_{TD} = h_{TD} x_T + n_{TD}$.

Finally, let us note that the fourth time-slot is not of interest in the detection process, as the bit transmitted in this time-slot is independent of all the others. So, it can be demodulated without considering previous received bits. However, the need of this time-slot to complete the overall communication is important to assess the network throughput of the system.

**C. Detection at Node $D$**

Upon reception of signals $y_{SD}$, $y_{RD}$, and $y_{TD}$ in time-slot one, two, and three, respectively, node $D$ can perform joint demodulation of $b_S$ and $b_T$. As mentioned above, $b_R$ is treated independently as the related packet is independent of the others. To avoid the analytical intractability and implementation complexity of the ML-optimum demodulator, we consider the sub-optimal, but asymptotically-tight (for high Signal–to–Noise–Ratio, SNR), Cooperative Maximum Ratio Combining (C–MRC) detector shown in [23] on top of this page [16], [22], where: i) $\lambda_R = \min \{\gamma_{SR}, \gamma_{RD}\}/\gamma_{RD}$ and $\lambda_T = \min \{\gamma_{ST}, \gamma_{TD}\}/\gamma_{TD}$ account for the reliability of the $S$–to–$R$ and $S$–to–$T$ links, respectively; and ii) $\gamma_{XY} = |h_{XY}|^2 (E_m/N_0)$ with $X$ and $Y$ being two generic nodes of the network. The derivation of (2) follows the same arguments as in [16], [22], and it is here omitted to avoid repetitions.

### III. PERFORMANCE ANALYSIS

The aim of this section is to estimate the performance of the detector in (2), by providing a closed-form expression of the ABEP for high–SNR. The ABEP of node $S$ and node $T$, i.e., $\text{ABEP}_S = \Pr \{b_S \neq \hat{b}_S^{(S)}\}$ and $\text{ABEP}_T = \Pr \{b_T \neq \hat{b}_T^{(T)}\}$, respectively, can be computed by using the methodology described in [19, Sec. IV]. In particular, we have:

$$
\text{ABEP}_X \leq \frac{1}{\text{card} \{C\}} \sum_{b_S = 0}^1 \sum_{b_T = 0}^1 \sum_{b_T = 0}^1 \sum_{b_T = 0}^1 \text{AEP}_X (e \rightarrow \tilde{e})
$$

where: i) $\text{AEP}_X (e \rightarrow \tilde{e}) = \text{AEP} (e \rightarrow \tilde{e}) \Delta (b_X, \tilde{b}_X)$; ii) $C = \{000, 010, 110, 101\}$ is the codebook of Scenario (d) in Fig. 2 which takes into account forwarding and NC operations performed at nodes $R$ and $T$. The generic element of $C$ is $e = [b_S, b_S, b_S \oplus b_T]$; iii) $\text{card} \{C\} = 4$ is the cardinality of $C$, i.e., the number of codewords $e$ in $C$; iv) $\text{AEP} (e \rightarrow \hat{e})$ is the Average Pairwise Error Probability (AEP) of the generic pair of codewords $e = [c_1, c_2, c_3] = [b_S, b_S, b_S \oplus b_T]$ and $\hat{e} = [\tilde{c}_1, \tilde{c}_2, \tilde{c}_3] = [b_S, b_S, b_S \oplus b_T]$ of the codebook, i.e., the probability of estimating $\hat{e}$ in (2), when, instead, $e$ has actually been transmitted, and $e$ and $\hat{e}$ are the only two codewords possibly being transmitted; and $v) \Delta (b_X, \tilde{b}_X) = 1^1 \Pr \{\cdot\}$ denotes probability.
1−\(\Delta (bx, \tilde{b}X)\), where \(\Delta (\cdot, \cdot)\) is the Kronecker delta function, i.e., \(\Delta (bx, \tilde{b}X) = 1\) if \(bx = \tilde{b}X\) and \(\Delta (bx, \tilde{b}X) = 0\) if \(bx \neq \tilde{b}X\). This function is used to include in the computation of \(\text{ABEP}_X\) only those APEPs which result in an error for the information bit of interest, i.e., \(X = S\) or \(X = T\) [19].

A. Computation of APEP \((c \rightarrow \tilde{c})\)

From (3), it follows that ABEP can be estimated if APEP \((c \rightarrow \tilde{c})\) is available in closed-form, where the average is over fading channel statistics and AWGN. In this section, we compute an asymptotically-tight formula for APEP \((c \rightarrow \tilde{c})\), which is accurate for high-SNR.

From (2), by definition, we have (4) on top of this page, where: i) \(\Lambda_{c} = \Lambda (b_S, b_T; b_S, b_T, \tilde{b}_S; \tilde{b}_S)\) and \(\Lambda_{\tilde{c}} = \Lambda (\tilde{b}_S, \tilde{b}_T; b_S, b_T, \tilde{b}_S; \tilde{b}_S);\) ii) \(E_X (\cdot)\) is the expectation operator computed over RV \(X;\) and iii) \((a)\) is obtained by using the total probability theorem and by conditionning upon possible decoding errors at nodes \(R\) and \(T\) [23]. Since demodulation outcomes at node \(R\) and \(T\) are independent, we have: i) \(Pr (\tilde{b}_S = b_S, \tilde{b}_T = b_T) = 1 - Q (\sqrt{2SD});\) ii) \(Pr (\tilde{b}_S = b_S, \tilde{b}_T \neq b_T) = 1 - Q (\sqrt{2SD});\) iii) \(Pr (\tilde{b}_S \neq b_S, \tilde{b}_T = b_T) = Q (\sqrt{2SD});\) and iv) \(Pr (\tilde{b}_S \neq b_S, \tilde{b}_T \neq b_T) = Q (\sqrt{2SD})\). These probabilities are functions to use in the expectation equation (5).

The next step is the computation of each conditional probability \(Pr (\Delta_{c, \tilde{c}} < 0 | \cdot)\). To this end, a closed-form expression of \(\Delta_{c, \tilde{c}}\) is needed. This can be obtained by substituting \(y_{SD}, y_{RD}, y_{TD}\) in (2), and through some algebraic manipulations. The final result is as follows:

\[
\Delta_{c, \tilde{c}} = \gamma_X d_S + 2\sqrt{SD} d_S \text{Re} \{\eta_{SD}\} + \lambda_T (\gamma_T d_R + 2\sqrt{TD} d_T \text{Re} \{\eta_{TD}\})
\]

where: i) \(\text{Re} \{\cdot\}\) is the real part operator; ii) \((\cdot)^*\) denotes complex conjugate; iii) \(j = \sqrt{-1}\) is the imaginary unit; iv) \(\phi_{XY}\) is the phase of the generic fading gain \(h_{XY}\), i.e., \(h_{XY} = |h_{XY}| \exp (j \phi_{XY})\); v) \(\eta_{XY} = \eta_{XY} \phi_{XY} / \sqrt{N_0}\) is the normalized AWGN for the generic X→Y link, which has zero mean and unit variance; vi) \(d_S = 2(\tilde{c}_1 - c_1) = 2 (\tilde{b}_S - b_S), d_R = 2(\tilde{c}_2 - c_2) = 2 (\tilde{b}_S - b_S), d_T = 2(\tilde{c}_3 - c_3) = 2 (\tilde{b}_S \oplus b_T - (b_S \oplus b_T));\) and vii) \(d_R = 2(1 - \tilde{b}_S) d_R, d_T = 2(1 - \tilde{b}_S) d_T\).

Finally, it is worth noticing that the expression given in (5) is useful whenever the conditioning on the bits estimated at node \(R\) and node \(T\) are. Only \(d_S\) and \(d_T\) change for different detection outcomes. To make this aspect more explicit, we use the notation \((X = R, X = T); i) d_X = d_X (ok) if \(\tilde{b}_X = b_S;\) and ii) \(d_X = d_X (ok) if \tilde{b}_X \neq b_S\).

To compute \(Pr (\Delta_{c, \tilde{c}} < 0 | \cdot)\), we exploit the Laplace inversion transform method in [24, Eq. (5)]:

\[
Pr (\Delta_{c, \tilde{c}} < 0 | \cdot) = \frac{1}{2\pi j} \int_{-j\infty}^{+j\infty} \mathcal{M}_{\Delta_{c, \tilde{c}}} (s) e^{s} ds
\]

with: i) \(\mathcal{M}_{\Delta_{c, \tilde{c}}} (s) = \mathcal{E}_b (\mathcal{Q}_X (\gamma_X, \mathcal{M}_{\Delta_{c, \tilde{c}}} (\cdot)))\) being the (two-sided) Moment Generating Function (MGF) of the conditional RV \(\Delta_{c, \tilde{c}}\). The average is computed over fading gains and AWGN of all the links \(X\)→\(D\) for \(X = \{S, R, T\}\); and ii) \(\delta\) being a real number such that the contour path of integration is in the region of convergence of \(\mathcal{M}_{\Delta_{c, \tilde{c}}} (\cdot)\).

Then, APEP \((c \rightarrow \tilde{c})\) can be obtained by substituting (6) in (4), by computing the expectation over fading statistics, AWGN, and by solving the inverse Laplace transform. In particular, since in this paper we are interested in high-SNR analysis, i.e., \(E_m / N_0 \rightarrow \infty\), an asymptotic expression of the MGF in (6) is needed [24, Eq. (12)]. Due to space constraints, in this paper we cannot provide all the details of the derivation. As an illustrative example, we provide a brief description of the main steps behind the computation of one addend in (4).

In particular, we focus our attention on the fourth addend in (4), which is denoted by APEP\(^{(4)}\) \((c \rightarrow \tilde{c})\). The reason is that this term is the most complicated to be computed. APEP\(^{(4)}\) \((c \rightarrow \tilde{c})\) in (4) can be written as shown in (7) on top of the next page, where: i) \(F_X (\cdot)\) is defined in (3) on top of the next page; ii) \(\bar{\Psi}_s (\cdot)\) is obtained by using the Craig’s representation of the Q-function [25]; iii) \(\bar{\Psi}_s (\cdot)\) is obtained by averaging over the AWGN with \(G (\cdot, \cdot)\) being defined in (9) on top of the next page; and iv) \(\bar{\Psi}_s (\cdot)\) is obtained by averaging over channel fading and using some simplifications that hold for high-SNR. In particular, \(\Psi_s (\cdot)\), \(\Psi_t (\cdot)\), and \(\Psi_2 (\cdot, \cdot)\) are defined in (10) on the top of the next page, where \(\gamma_X = 2\sigma_X (E_m / N_0)\) for the generic pair of nodes \(X\) and \(Y\). Note that, for \(X = R\) and \(X = T\), \(d_X \neq 0 \Rightarrow d_X \neq 0\). Let us consider the most general case with \(d_S \neq 0, d_R \neq 0,\) and \(d_T \neq 0\). Both integrals in the brackets in (10) can be computed in closed-form with the help of [25, Eq. (5A.9)]. Thus, APEP\(^{(4)}\) \((c \rightarrow \tilde{c})\) simplifies as follows:

\[
\text{APEP}_{(c \rightarrow \tilde{c})} = \frac{\mathcal{I}_4 (\phi_{XY}^{(ok)}, \delta_{XY}^{(ok)})}{4\gamma_{SD} \gamma_{RD} \gamma_{TD} \sqrt{N_0} \sqrt{SD} d_S \delta_{XY}^{(ok)} d_T^{(ok)}}
\]

where \(\mathcal{I}_4 (\cdot, \cdot)\) is defined in (12) on top of the next page.

Some important considerations are worth being made about APEP\(^{(4)}\) \((c \rightarrow \tilde{c})\) in (11). First, we notice that the asymptotic behavior of the APEP is clearly shown, and, for the considered case study, a diversity order equal to three is obtained [8].

Second, the integral \(\mathcal{I}_4 (\cdot, \cdot)\) can be computed, either analytically or numerically, by using one of the many methods described in [24]. Finally, we would like to mention that the case study investigated in this section, i.e., APEP\(^{(4)}\) \((c \rightarrow \tilde{c})\), is the most complicated addend, as it is the only term involving the product of two Q-functions. All the other cases are much simpler to be computed, and all integrals similar to \(\mathcal{I}_4 (\cdot, \cdot)\) in
Let us consider the 3–node network topology. The ABEP determine both diversity and coding gain. The accuracy of as a function of far as node Node \( S \) has no gain from cooperation, and the diversity order to the non–cooperative scenario, and, in some cases, NC might also be harmful. To the best of the authors knowledge, this important behavior has never been reported in the open technical literature [4]. Similar comments apply to the 4–node network topology. In particular, we notice that node \( S \) has a diversity order that depends on the number of relay nodes that do not perform NC but just forward the received packets.

Finally, we would like to emphasize that, unlike state–of–the–art performance analysis of cooperative networks (see [15], [16], [20] for further comments), our analysis encompasses a very accurate estimation of the coding gain. This is instrumental to clearly assess diversity and coding trade–off summarized in Table IV.

### IV. PERFORMANCE COMPARISON: IS NC USEFUL?

The aim of this section is to compare the performance of the different scenarios and network topologies shown in Fig. 1 and Fig. 2. For all cases of interest, the methodology described in Section III is used to compute the ABEP. In particular, (3) is applied for all possible codewords of the codebook. The final results are summarized in Table I by assuming, for a fair comparison, the total energy constraint mentioned in Section III, and the theoretical findings Section IV.

Furthermore, since we are interested in high–SNR analysis, Table I shows only the dominant terms in (3), i.e., those APEPs having the slowest decaying behavior as a function of \( E_m/N_0 \to \infty \) [19]. In fact, these terms determine both diversity and coding gain. The accuracy of the frameworks shown in Table I is validated in Section V through Monte Carlo simulations.

Important considerations can be drawn from this analysis. Let us consider the 3–node network topology. The ABEP of Scenario (b) shows that node \( S \) can exploit distributed diversity to improve the diversity gain, but the price to pay is a performance degradation for node \( R \), whose ABEP is worse than in the non–cooperative case, i.e., Scenario (a). Very interestingly, we notice that the network–coded scenario, i.e., Scenario (c), is the worst one in terms of performance. Node \( S \) has no gain from cooperation, and the diversity order is equal to one. Furthermore, and very surprisingly, node \( S \) has the same ABEP as in the non–cooperative case. In other words, there is neither power nor diversity gain. As far as node \( R \) is concerned, the situation is even worse: the ABEP is worse than the non–cooperative case. Also, we notice that this performance penalty depends only on part in decoding errors on the \( S \)–to–\( R \) link. In fact, even assuming \( \gamma_{SR} \to \infty \), i.e., no decoding errors at node \( R \), the ABEP is worse because of performing NC. In conclusion, unlike [16], [19]–[21] where it shown that NC is beneficial in cooperative networks when some nodes act only as relays and have no data to transmit, Table I points out that, if the relay nodes have their own data to transmit, NC introduces no gain when compared to the non–cooperative scenario, and, in some cases, NC might also be harmful.
TABLE I
ABEP for high-SNR ($k_t \approx 0.4853$ is obtained by computing terms like $J_1\left(q^{(1)}_m, \sigma^2\right)$ in (11) by using (24, Eq. (10)), and $k_2 = \left(525 + 11\sqrt{5}\right) / 800$.

| Scenario     | ABEP_σ | ABEP_R | ABEP_T |
|--------------|--------|--------|--------|
| 3-Node Network (a) | $\left(1/4\right) \gamma^{-1}_D$ | $\left(1/4\right) \gamma^{-1}_D$ | - |
| 3-Node Network (b) | $\left(3/8\right) \gamma^{-1}_D \gamma^{-1}_R + \left(45 + \sqrt{7}\right) / 160 \gamma^{-1}_D \gamma^{-1}_S$ | $\left(1/2\right) \gamma^{-1}_D$ | - |
| 3-Node Network (c) | $\left(1/4\right) \gamma^{-1}_D$ | $\left(1/4\right) \gamma^{-1}_D + \left(1/4\right) \gamma^{-1}_S$ | - |
| 4-Node Network (a) | $\left(1/4\right) \gamma^{-1}_D$ | $\left(1/4\right) \gamma^{-1}_D$ | $\left(1/4\right) \gamma^{-1}_D$ |
| 4-Node Network (b) | $\left(5/8\right) \gamma^{-1}_D \gamma^{-1}_R \gamma^{-1}_T + \left(45 + \sqrt{7}\right) / 160 \gamma^{-1}_D \gamma^{-1}_S$ | $\left(1/2\right) \gamma^{-1}_D$ | - |
| 4-Node Network (c) | $\left(1/4\right) \gamma^{-1}_D$ | $\left(1/4\right) \gamma^{-1}_D + \left(1/4\right) \gamma^{-1}_S$ | - |
| 4-Node Network (d) | $\left(3/8\right) \gamma^{-1}_D \gamma^{-1}_R + \left(45 + \sqrt{7}\right) / 160 \gamma^{-1}_D \gamma^{-1}_S$ | $\left(1/4\right) \gamma^{-1}_D + \left(1/4\right) \gamma^{-1}_T$ | - |

Fig. 3. ABEP against $E_{n0}/N_0$ for the 3-node network topology in Fig. 1. Solid lines show the analytical framework and markers show Monte Carlo simulations. i) i.i.d. fading with $\sigma^2_n = 1$; and ii) $\gamma_0 = 2\sigma^2_n (E_{n0}/N_0)$. ABEP_σ and ABEP_R of Scenario (a) are given by the single-hop bound.

Fig. 4. ABEP against $E_{n0}/N_0$ for the 4-node network topology in Fig. 1. Solid lines show the analytical framework and markers show Monte Carlo simulations. i) i.i.d. fading with $\sigma^2_n = 1$; and ii) $\gamma_0 = 2\sigma^2_n (E_{n0}/N_0)$. ABEP_σ, ABEP_R, and ABEP_T of Scenario (a) are given by the single-hop bound.

ACKNOWLEDGMENT
This work is supported, in part, by the research projects “GREENET” (PTN–GA–2010–264759), “WSN4QoL” (IAPP–GA–2011–286047), and the Lifelong Learning Programme ( LLP) – ERASMUS Placement.

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