Abstract—We analyze coding for energy efficiency in relay channels at a fixed source rate. We first propose a half-duplex decode-forward coding scheme for the Gaussian relay channel. We then derive three optimal sets of power allocation, which respectively minimize the network, the relay and the source energy consumption. These optimal power allocations are given in closed-form, which have so far remained implicit for maximum-rate schemes. Moreover, analysis shows that minimizing the network energy consumption at a given rate is not equivalent to maximizing the rate given energy, since it only covers part of all rates achievable by decode-forward. We thus combine the optimized schemes for network and relay energy consumptions into a generalized one, which then covers all achievable rates. This generalized scheme is not only energy-optimal for the desired source rate but also rate-optimal for the consumed energy. The results also give a detailed understanding of the power consumption regimes and allow a comprehensive description of the optimal message coding and resource allocation for each desired source rate and channel realization. Finally, we simulate the proposed schemes in a realistic environment, considering path-loss and shadowing as modelled in the 3GPP standard. Significant energy gain can be obtained over both direct and two-hop transmissions, particularly when the source is far from relay and destination.

I. INTRODUCTION

As a key feature of future wireless systems, relaying has attracted significant attention in recent years and a rich literature has contributed to analyzing the capacity of relay channels. Upper and lower bounds were first established in [1], and derived for AWGN relay channels considering energy-per-bit in [2]. More recently, bounds have been extended to multiple relays in [3] and MIMO channels in [4] [5]. Considering the three-node networks, power allocation maximizing the source rate at a given power has been extensively analyzed in [6] [7].

In practice, however, user applications are mainly associated with a fixed minimum rate, defined as a Quality-of-Service feature [8], and run on power-limited devices. This limitation has spurred new analysis and designs for energy efficiency. For example, bounds on the minimum energy-per-bit have been established in [2] and power allocations for energy efficiency of the relay channel have been derived in [9] [10]. Despite of this interest, energy efficiency analysis is mostly based on power allocations optimized for rate as in [10]. Coding and optimization directly for energy remain scarce. One reason is that the two problems of maximizing rate at a given network power and minimizing network power at a given rate are often thought to be equivalent. Nevertheless, neither optimization problem is fully characterized. Power allocations optimized for rate remain implicit without closed-form solutions, while power allocations for energy remain scarce and are mainly based on unrealistic constraints, such as full-duplex transmissions [9] or no individual power constraints [9] [10].

In this paper, we explore energy optimization for the three-node relay channel in Gaussian noise environments, considering Gaussian signaling as typically used in such analysis to define upper-bounds for the system performance [11]. We first propose a comprehensive half-duplex coding scheme based on time division for fading channels with Gaussian noise. We then explicitly optimize its resource allocation for energy efficiency at a desired source rate and individual node power constraints. We consider three objectives: the network, the relay and the source energy consumption. We next combine the proposed schemes into a generalized one that maximizes the energy efficiency of the Gaussian relay channel and analyze its performance as a function of both the source rate and the mobile user location. The contributions are threefold:

• We show that maximizing the source rate and minimizing the network energy consumption are not equivalent as often believed. Specifically, there are different power-optimized regimes that cannot be uncovered by maximizing rates.
• We highlight the different regimes of coding technique and optimal power allocation (full or partial decode-forward, with or without beamforming), which constitute the lower bound on the capacity of the Gaussian relay channel by using decode-forward.
• We provide the optimal power allocation in closed-form.

The paper is organized as follows. The new half-duplex coding scheme is analyzed for the Gaussian relay channel in Section II. We optimize the resource allocation for network, relay and source energy efficiency in Section III and build the generalized scheme in Section IV. Section V presents the performance analysis and Section VI concludes this paper.
transmit power. Denote node allocates to each message a portion of its available power $P$. This scheme is depicted in Figure 1.

$\rho_m$ end of the second phase, whereas $m_R$ divides the initial message into two parts

message splitting. It divides the initial message into two parts

in the half-duplex partial decode-forward scheme presented in

the direct link, the source-to-relay link and the relay-to-

destination link, as depicted in Figure [1] We assume that

can occur. 

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the first and second phase

where at each time either $Y_r$ and $Y_1$ occur together or only $Y_2$ can occur.

B. Coding scheme for the Gaussian relay channel

We consider Gaussian signaling and superposition coding as in

the half-duplex partial decode-forward scheme presented in

our conference version [12] for the discrete memoryless and

Gaussian channels. For cooperative transmission, superposi-
tion modulation benefits from a renewed interest for potential

implementation. As shown in [13], this strategy outperforms

other techniques, such as bit-interleaved coded modulation, at

even lower receiver complexity. 

We describe the coding scheme as follows. To send a

message $m$ of rate $R$ to the destination, the source performs

message splitting. It divides the initial message into two parts

$(m_d, m_r)$, with rates $R_d$ and $R_r$ respectively, where $R_d+R_r = R$. Both are encoded at the source using superposition coding. The message $m_d$ is directly decoded by the destination at the end of the second phase, whereas $m_r$ is intended to be relayed. This scheme is depicted in Figure [1]

The source and the relay have individual power constraints $P_s$ and $P_r$ within the same bandwidth. At each phase, each node allocates to each message a portion of its available transmit power. Denote $\eta_1$ and $\eta_2$ as the portion of source power $P_s$ allocated to $m_d$ in the first and second phase respectively, similarly, $\rho_1$ and $\rho_2$ as the portion for $m_r$. Denote $\rho_r$ as the portion of the relay power $P_r$ used to forward $\hat{m}_r$ to the destination. We consider transmit power constraint at each node such that

$$P_s^{(c)} = \theta (\eta_1 + \rho_1) P_s + \tilde{\theta} (\eta_2 + \rho_2) P_s \leq P_s$$

$$P_r^{(c)} = \tilde{\theta} \rho_r P_r \leq P_r$$

where superscript $(c)$ refers to the consumed power.

Applying the scheme to the Gaussian relay channel, the transmit signal can be written in two separate phases as the channel in [1] as follows

$$\text{Phase 1: } X_1 = \sqrt{\rho_1 P_s} U + \sqrt{\eta_1} P_s V$$

$$\text{Phase 2: } X_r = \sqrt{\rho_r P_r} U ; X_2 = \sqrt{\rho_2 P_s} U + \sqrt{\eta_2} P_s V$$

where $U(m_r) \sim N(0,1)$ and $V(m_d) \sim N(0,1)$ are independent Gaussian signals that carry messages $m_r$ and $m_d$ respectively. Note that $U$ appears in both $X_r$ and $X_1$, which allows a beamforming gain at the destination.

Joint decoding is considered at the receivers. At the end of Phase 1, the relay decodes $m_r$, i.e. chooses the unique $\hat{m}_r$ for which $U$ and $Y_r$ are typical. At the end of Phase 2, the destination jointly decodes $(m_d, m_r)$ by choosing the unique $(\hat{m}_d, \hat{m}_r)$ for which $U$ and $X_1$ are jointly typical with $Y_1$, and simultaneously $U$, $X_2$ and $X_r$ are jointly typical with $Y_2$. More details on decoding can be found in our previous work [12]. Note that instead of joint typicality, maximum likelihood decoding can be used to achieve the same rate.

For Gaussian relay channels, all rates satisfying constraints in Eq [4] at the top of the page are achievable. In this equation, $\theta \in [0,1]$, and power set $(\eta_1, \rho_1, \eta_2, \rho_2, \rho_r)$ is non-negative and satisfies constraint (2). Both rate constraints are results of the decoding at the relay over the first phase and joint decoding at the destination simultaneously over both phases. The first constraint captures the coherent transmission between source and relay in the second phase, whereas the second constraint comes from the decoding of $m_r$ at the relay and decoding of $m_d$ at the destination. We refer to [12] for the general proof in the discrete memoryless channel case. This coding scheme includes as special cases direct transmissions, two-hop relaying, as well as the maximum-rate scheme proposed in [6] [7].

\[ R \leq \theta \log_2 \left( 1 + \frac{\eta_1 P_s |h_d|^2}{N} \right) + \tilde{\theta} \log_2 \left( 1 + \frac{(\eta_2 + \rho_2) P_s |h_d|^2 + \rho_r P_r |h_r|^2 + 2\sqrt{P_s |h_d|^2 P_r |h_r|^2 |\rho_2 \rho_r|}}{N} \right) = I_1 \]

\[ R \leq \theta \log_2 \left( 1 + \frac{\rho_1 P_s |h_d|^2}{N} \right) + \tilde{\theta} \log_2 \left( 1 + \frac{\eta_1 P_s |h_d|^2}{N} \right) + \tilde{\theta} \log_2 \left( 1 + \frac{\eta_2 P_r |h_r|^2}{N} \right) = I_2 \]
C. Energy-optimization problem

We propose to optimize the above coding scheme for energy efficiency in the Gaussian relay channel at a given source rate. We consider three cases: the network, the relay and the source power consumptions. Specifically, we look for the optimal set of power allocation $(\rho_1, \eta_1, \rho_2, \eta_2, \rho_r)$ solving the following general problem:

$$
\min \omega_s \left[ \theta (\eta_1 + \rho_1) P_s + \theta (\eta_2 + \rho_2) P_s \right] + \omega_r \theta \rho_r P_r \quad (5)
$$

s.t.

$$
I_1 \geq R ; \quad I_2 \geq R
$$

$$
P_s^{(c)} \leq P_s ; \quad P_r^{(c)} \leq P_r
$$

where $\omega_s$ and $\omega_r$ are either 0 or 1, depending on the targeted optimization. The two rate constraints ensure achievability of the source rate, given two individual power constraints.

Optimization for Gaussian signaling in Gaussian channel as formulated in (5) provides an upper-bound on the performance of practical systems. It defines the limit of what is possible in practice, as well as gives an insight of how a practical scheme should be structured. Especially, it is found in [14] that the system that emulates most closely the structure of signals defined by information theory also reaches the highest performance. The authors of [14, 15] analyze the maximum rates allowed by information theory for the multiple-access Gaussian channel using cooperation, as well as implementation in a CDMA system. Other analysis is provided in [16] for the interference Gaussian channel in a tri-sectored OFDMA network. In this reference, an interference mitigation scheme with power allocation is proposed for Gaussian signaling and then is simulated for various practical modulations and coding rates. The practical system approaches the channel capacity with Gaussian signaling within 1.2 bit per channel use. These examples of existing works illustrate that optimization for Gaussian signaling is a valid approach for determining optimal power allocation for practical implementation.

III. THREE ENERGY EFFICIENT SCHEMES FOR THE RELAY CHANNEL

In this section, we explore energy optimization for the relay channel considering three objectives: the network, the relay and the source energy consumption. Then, we compare the maximum achievable rate of each optimized scheme.

A. Network energy optimal set of power allocation (N-EE)

In this optimization, we consider the total source and relay power consumption during both transmission phases. To minimize the total consumption while maintaining a desired source rate, we analyze the following problem $(\omega_s = 1, \omega_r = 1)$, denoted as N-EE:

$$
\min \left[ \theta (\eta_1 + \rho_1) P_s + \theta (\eta_2 + \rho_2) P_s \right] + \theta \rho_r P_r \quad (6)
$$

s.t.

$$
I_1 \geq R ; \quad I_2 \geq R
$$

$$
P_s^{(c)} \leq P_s ; \quad P_r^{(c)} \leq P_r
$$

The optimal power allocation depends on the desired rate. If the source rate is higher than a certain threshold $R^{(n)}$, then partial decode-forward is applied (sub-scheme $A^{(n)}$), as detailed in Proposition [1]. Otherwise, the source does not split its message and uses full decode-forward (sub-scheme $B^{(n)}$), with power allocation as given in Proposition [2]. Note that the power consumption of N-EE is a continuous function of the source rate, even at $R^{(n)}$. This result is summarized in Algorithm [3] and the proof can be found in Appendix A.

**Proposition 1.** In sub-scheme $A^{(n)}$, the source splits its message into two parts $m_r$ and $m_d$.

- In Phase 1, the source sends $m_r$ with power $\rho_1^* P_s$ and rate $R_r = \theta \log_2 \left( 1 + \frac{\rho_1^* P_s |h_s|^2}{N} \right)$. The relay decodes as $\tilde{m}_r$.
- In Phase 2, the relay sends $\tilde{m}_r$ with power $\rho_r^* P_r$ and the source sends $(m_r, m_d)$ with power $(\rho_2^* P_s, \eta_2^* P_s)$. The rate of $m_d$ is $R_d = R - R_r$.

To compute the optimal power allocation set for $A^{(n)}$, $\rho_1^*$ is first found numerically by solving Eq. [7] of Algorithm [3]. Next, $(\rho_r^*, \rho_2^*, \eta_1^*, \eta_2^*)$ are deduced from $\rho_1^*$ as follows:

$$
\eta_1^* = 0 ; \quad \rho_1^* = \frac{N |h_s|^2}{P_r (|h_r|^2 + |h_d|^2)^2} \left( \frac{2^{R/\theta}}{N} \right)^{\theta/\theta} - \frac{2^{R/\theta}}{N} \left( \frac{2^{R/\theta}}{N} \right)^{\theta/\theta}
$$

$$
\eta_2^* = \frac{2^{R/\theta}}{N} \left( \frac{2^{R/\theta}}{N} \right)^{\theta/\theta} - \frac{N |h_d|^2}{P_s |h_d|^2} ; \quad \rho_2^* = \frac{P_s |h_d|^2}{P_s |h_r|^2} \rho_r^*
$$

Regarding the computation of optimal allocation, solving Eq. [7] of Algorithm [3] in terms of $v_s$, rather than in terms of $\rho_1^*$, reduces to solving a polynomial equation and is sufficient to deduce the whole allocation. This shows that computing the optimal allocation is relatively simple, especially in case
of equal time division (θ = ̂θ = 1/2) which is widely used in practical systems.

As shown in Algorithm I partial decode-forward is applied as long as \( R^{(n)} \leq R \leq R^{(n)}_{\text{max}} \). The lower bound \( R^{(n)} \) comes from the non-negativity constraint on \( η_2 \). The upper bound \( R^{(n)}_{\text{max}} \) corresponds to the maximum rate that is achievable given the power constraints defined in (2). We will come back to this upper bound in Section III-D. Note that, when the direct link is close to 0, the maximum feasible rate for \( m_d \) goes to zero. Due to power constraints, outage can occur at some rate \( R \leq R^{(n)} \), such that sub-scheme \( A^{(n)} \) may not exist. When \( R \leq R^{(n)} \), we apply sub-scheme \( B^{(n)} \) as discussed next.

**Proposition 2.** In sub-scheme \( B^{(n)} \), the source does not split its message \( m \).
- In Phase 1, the source sends \( m \) with power \( ρ_1^1P_s \) and the relay decodes as \( m \).
- In Phase 2, the relay sends \( m \) with power \( ρ_1^2P_r \) and the source sends \( m \) with power \( ρ_2^2P_s \).

The optimal power allocation for sub-scheme \( B^{(n)} \) is

\[
η_1^n = η_2^n = 0
\]

\[
ρ_1^1 = \left( \frac{2R/θ - 1}{P_s|h_s|^2} \right) \cdot \frac{N}{P_s|h_s|^2}; \quad ρ_2^2 = \frac{P_r|h_d|^2}{P_s|h_r|^2} \cdot \frac{P_s}{P_r}
\]

\[
ρ_1^2 = \frac{N|h_s|^2}{P_r (|h_s|^2 + |h_d|^2)} \cdot \left( \frac{2R/θ}{\left(1 + \frac{ρ_1^1P_s|h_s|^2}{N}\right)^{θ/2}} - 1 \right).
\]

\( B^{(n)} \) is applied when the source rate is low. In this case, full decode-forward is energy-efficient and the whole message benefits from the beamforming gain between the source and the relay.

As an example, Figure 2 illustrates the power consumptions of the above two sub-schemes, which differ mainly in the second phase. Thus, the figure plots the set of optimal power allocation and the total energy consumed by the source and the relay during this phase, as a function of the source rate. When the source rate is not achievable given power constraints, an outage occurs, which is depicted by a cut-off in the curve.

**Algorithm II**

**Optimal scheme for relay energy efficiency, R-EE, solving (8)**

**B. Relay energy optimal set of power allocation (R-EE)**

Next, we minimize the relay power consumption only and allow the source to consume up to \( P_s \) during both phases of the transmission. This scheme is of particular interest in networks where the relay is shared and has to serve many users, or where the relay has its own data to send. Thus, we analyze the following optimization problem, denoted as R-EE:

\[
\min \bar{ρ}_r P_r
\]

s.t.

\[
I_3 \geq R; \quad I_2 \geq R; \quad P_s^{(c)} \leq P_s; \quad P_r^{(c)} \leq P_r
\]

Once again, the optimal power allocation depends on the desired rate. Direct transmission (sub-scheme \( C^{(r)} \)) is used whenever feasible, i.e. as long as the source rate is under the capacity of the direct link, denoted as \( R^{(r)}_1 \). Note that direct transmissions can be optimal for relay energy but not for source or network energy. Indeed, the optimization problem considered here is relaxed in terms of the source consumption. This implies that some sub-schemes can be optimal for R-EE but not for N-EE.

If direct transmission is not feasible, the optimal scheme for relay energy uses, similarly to N-EE, either partial decode-forward (sub-scheme \( A^{(r)} \)) or full decode-forward (sub-scheme \( B^{(r)} \)), depending on a certain threshold \( R^{(r)}_2 \). This result is summarized in Algorithm II and the proof can be found in Appendix B.

In the following propositions, we present each sub-scheme and associated power allocation going from low to high rates.
A scheme which aims at minimizing the relay energy consumption should always give priority to direct transmissions, such that, if possible, the relay is not used at all. Thus, as long as the direct link is strong enough to support the source rate, R-EE uses sub-scheme \( C^{(r)} \).

**Proposition 3.** Sub-scheme \( C^{(r)} \) is direct transmission. The optimal power allocation is as follows.

\[
\eta_1^* = \eta_2^* = (2^R - 1) \frac{N}{P_s|h_d|^2}; \quad \rho_1^* = \rho_2^* = \rho_r^* = 0.
\]

If the source rate increases above the capacity of the direct link, relaying is required and either sub-scheme \( A^{(r)} \) or \( B^{(r)} \) is applied, as discussed next. These sub-schemes are similar to those in N-EE but with different power allocation. In the case of R-EE, the source consumes all its available power. If \( R \leq R_2^{(r)} \) as in Algorithm 1, we apply sub-scheme \( B^{(r)} \).

**Proposition 4.** In sub-scheme \( B^{(r)} \), the source uses full decode-forward.

- **In Phase 1**, the source sends \( m \) with power \( \rho_1^* P_s \). The relay decodes as \( \tilde{m} \).
- **In Phase 2**, the relay sends \( \tilde{m} \) with power \( \rho_2^* P_r \) and the source sends \( m \) again, with power \( \rho_r^* P_s \).

The optimal power allocation for sub-scheme \( B^{(r)} \) is

\[
\eta_1^* = \eta_2^* = 0
\]

\[
\rho_1^* = \left(\frac{2R}{\theta} - 1\right) \frac{N}{P_s|h_d|^2}; \quad \rho_2^* = \frac{1 - \rho_1^* \theta}{\theta};
\]

\[
\rho_r^* = \frac{N}{P_r|h_r|^2} \left( \sqrt{g_3(\eta_1^*)} - \sqrt{P_s|h_d|^2 \rho_2^*} \right)^2.
\]

Note that sub-scheme \( B^{(r)} \) only exists if the channel gains are such that \( R_1^{(r)} < R_2^{(r)} \). This condition essentially depends on the strength of the direct link. When \( R_1^{(r)} \geq R_2^{(r)} \), sub-scheme \( A^{(r)} \) is directly applied.

When the source rate increases even further to be above \( R_2^{(r)} \), message splitting is required and sub-scheme \( A^{(r)} \) is applied.

**Proposition 5.** In sub-scheme \( A^{(r)} \), the source splits its message into two parts \( m_r \) and \( m_d \).

- **In Phase 1**, the source sends \( m_r \) with power \( \rho_1^* P_s \) and rate \( R_r = \theta \log_2 \left( 1 + \frac{\rho_1^* P_s|h_r|^2}{N} \right) \). The relay decodes as \( \tilde{m} \).
- **In Phase 2**, the relay sends \( \tilde{m} \) with power \( \rho_2^* P_r \) and the source sends \( (m_r, m_d) \) with power \( (\rho_2^* P_s, \eta_2^* P_s) \). The rate of \( m_d \) is \( R_d = R - R_r \).

To compute the optimal power allocation set for sub-scheme \( A^{(r)} \), \( \rho_1^* \) is first found numerically by solving Eq. 9 of Algorithm 1. Next, \( g_3(\rho_1^*) \) is computed using \( 10 \) and \( (\rho_r^*, \rho_2^*, \eta_1^*, \eta_2^*) \) are deduced from \( \rho_1^* \) and \( g_3(\rho_1^*) \) as follows:

\[
\eta_1^* = \frac{1 - \rho_1^* \theta}{\theta} - \eta_2^*;
\]

\[
\rho_2^* = \frac{N}{P_r|h_r|^2} \left( \sqrt{g_3(\rho_1^*)} - \sqrt{P_s|h_d|^2 \rho_2^*} \right)^2.
\]

If \( \eta_1^* \) and \( \rho_1^* \) as in Proposition 3 do not satisfy the power constraints in 2, then the desired source rate cannot be achieved and outage is declared. For example and similarly to \( A^{(n)} \), when the direct link is very weak, sub-scheme \( A^{(r)} \) may not exist.

Figure 3 illustrates the above sub-schemes and plots the total energy consumption during the second phase as a function of the source rate. Somewhat unexpectedly, the energy consumed during this second phase is decreasing in \( R \) when applying sub-scheme \( B^{(r)} \). Since the destination decodes the message using the received signals during both phases, the optimization shows that it is relay energy-optimal for the source to allocate more energy during the first phase. Nevertheless, the total energy consumed during both phases is strictly increasing with rate, as expected.

**C. Source energy optimal set of power allocation (S-EE)**

As the last problem, we minimize the source power consumption alone, which is relevant to networks in which the relay is not power critical. We thus allow the relay to consume up to \( P_r \) during both transmission phases. We analyze the following optimization problem, denoted as S-EE:

\[
\min \theta (\eta_1 + \rho_1) P_s + \theta (\eta_2 + \rho_2) P_s
\]

s.t.

\[
I_1 \geq R; \quad I_2 \geq R; \quad \eta_1 \leq \bar{P}_s; \quad \rho_1 \leq P_r
\]

The optimal scheme for source energy efficiency is a function of the source rate and is composed of four sub-schemes: two-hop relaying (\( D^{(s)} \)), decode-forward (\( B^{(s)} \)), partial decode-forward with beamforming (\( A^{(s)} \)) and without beamforming (\( C^{(s)} \)). Note that two-hop relaying and partial decode-forward without beamforming are optimal only for source energy but not for relay or network energy. These sub-schemes are applied depending on the source rate compared with two thresholds \( R_1^{(s)} \) and \( R_2^{(s)} \), which are closely related.
to the non-negativity constraints on \( \rho_2 \) and \( \eta_2 \). This result is summarized in Algorithm III and the proof can be found in Appendix C. Note that the functions \( g_2 \) and \( g_3 \) of Algorithm III are the same in Algorithm II.

In the following propositions, we present each sub-scheme going from low to high rates and refer to Figure 4 for an illustration. For very low rates, it is energy-efficient for the source to apply two-hop relaying, such that the source only consumes energy during the first phase. However, contrary to simple routing, the destination is listening during both transmission phases. This corresponds to sub-scheme \( D^{(s)} \).

**Proposition 6.** Sub-scheme \( D^{(s)} \) is two-hop relaying with destination simultaneously decoding over both phases.

- **In Phase 1**, the source sends \( m_r \) and the relay decodes as \( \tilde{m}_r \).
- **In Phase 2**, the relay sends \( \tilde{m}_r \) and the source is silent. The optimal power allocation is as follows:

\[
\eta^{\rho_1}_1 = \eta^{\rho_2}_1 = \rho_2 = 0; \quad \rho_1 = \left( 2R/\theta - 1 \right) \frac{N}{P_s|h_s|^2} \rho_1^\ast = \frac{N}{P_r|h_r|^2} g_3(\rho_1^\ast)
\]

where \( g_3 \) is defined by (13) of Algorithm III.

If the rate increases, two-hop relaying is no longer energy-efficient for the source. In this case, either sub-scheme \( C^{(s)} \) or \( B^{(s)} \) is applied, depending on which of the SR- or RD-link is limiting. If the SR-link is the bottleneck of the network, the second rate constraint \( R \leq I_2 \) is the limiting one and the constraint on the relay allocation is relaxed. In this case, \( R_C \leq R_B \), as defined in Algorithm III and sub-scheme \( C^{(s)} \) is applied.

**Proposition 7.** Sub-scheme \( C^{(s)} \) uses partial decode-forward without beamforming.

- **In Phase 1**, the source sends \( m_r \) and the relay sends \( \tilde{m}_r \).
- **In Phase 2**, the source sends \( m_{rd} \) and the relay sends \( \tilde{m}_r \).

**Optimal scheme for source energy efficiency, S-EE, solving** (11) **to the non-negativity constraints on** \( \rho_2 \) **and** \( \eta_2 \). This result is summarized in Algorithm III and the proof can be found in Appendix C. Note that the functions \( g_2 \) and \( g_3 \) of Algorithm III are the same in Algorithm II.

In the following propositions, we present each sub-scheme going from low to high rates and refer to Figure 4 for an illustration. For very low rates, it is energy-efficient for the source to apply two-hop relaying, such that the source only consumes energy during the first phase. However, contrary to simple routing, the destination is listening during both transmission phases. This corresponds to sub-scheme \( D^{(s)} \).

**Proposition 6.** Sub-scheme \( D^{(s)} \) is two-hop relaying with
only sending new information \((m_d)\) during the second phase, and does not require beamforming (which is possible if both the source and the relay transmit in the second phase). Also note that, as the RD-link is non limiting, the relay does not necessarily use all its available power to transmit \(\tilde{m}_r\).

On the other hand, if the RD-link is the bottleneck of the network \((R_D \leq R_C)\), the relay cannot support the transmission of \(\tilde{m}_r\) alone, even by transmitting with full power. In this case, the source should repeat the message \(m_r\) during Phase 2, such that beamforming gain is obtained. Then, \(B^{(s)}\) is applied as follows.

**Proposition 8.** Sub-scheme \(B^{(s)}\) uses full decode-forward.

- In Phase 1, the source sends \(m\), the relay decodes as \(\tilde{m}\).
- In Phase 2, the relay sends \(\tilde{m}\) and the source continues to send \(m\).

The optimal power allocation for sub-scheme \(B^{(s)}\) is

\[
\rho_r^1 = \frac{1}{\theta} \quad \rho_s^1 = \left(2^{\frac{R}{\theta}} - 1\right) \frac{N}{P_s |h_d|^2} \quad \eta_1 = \eta_2 = 0
\]

\[
\rho_r^2 = \frac{N}{P_s |h_d|^2} \left(\sqrt{g_3(\rho_2^1)} - \sqrt{\frac{P_r |h_r|^2}{\theta N}}\right)^2
\]

Finally, if the source rate is very high, such that \(R \geq \max \left\{R^{(s)}_1, R^{(s)}_2\right\}\), partial decode-forward with beamforming is required and sub-scheme \(A^{(s)}\) is applied.

**Proposition 9.** In sub-scheme \(A^{(s)}\), the source splits its message into two parts \(m_r\) and \(m_d\).

- In Phase 1, the source sends \(m_r\) and the relay decodes \(m_r\) as \(\tilde{m}_r\).
- In Phase 2, the relay sends \(\tilde{m}_r\) and the source sends \((m_r, m_d)\).

To compute the optimal power allocation set, \(\rho^*_1\) is first found numerically by solving (12) of Algorithm [17]. Next, \(g_3(\rho^*_1)\) is computed using (13) and \((\rho^*_2, \eta_1^*, \eta_2^*)\) are deduced from \(\rho^*_1\) and \(g_3(\rho^*_1)\) as follows:

\[
\eta_2^* = \left(\frac{2^{\frac{R}{\theta}}}{1 + \frac{\rho^*_1 P_r |h_d|^2}{N}}\right)^{\frac{1}{\theta}} - 1 \quad \rho^*_2 = \frac{1}{\theta} \quad \rho^*_1 = \frac{N}{P_s |h_d|^2} \left(\sqrt{g_3(\rho^*_1)} - \sqrt{\frac{P_r |h_r|^2}{\theta N}}\right)^2
\]

This sub-scheme is similar to \(A^{(n)}\) and \(A^{(r)}\), but with different power allocation. Even if the relay already transmits with full power, the RD-link is still too weak to support the transmission of \(m_r\) alone. Thus, the source repeats the message \(m_r\) during the second phase to obtain beamforming gain. Moreover, contrary to \(B^{(s)}\), the SR-link is also too weak to support the transmission of the whole source message. Thus, part of the source message \((m_d)\) has also to be transmitted via the direct link only.

**Existence of the sub-schemes:** Considering the optimized scheme as function of the source rate, S-EE is composed of either \((D^{(s)}, B^{(s)}, A^{(s)})\) or \((D^{(s)}, C^{(s)}, A^{(s)})\), in that order, depending on which link is the bottleneck of the network. Depending on the power constraints and channel gains, S-EE can also be composed of fewer than three sub-schemes and outage can occur with any sub-scheme \(D^{(s)}, C^{(s)}, B^{(s)}\) or \(A^{(s)}\). Specifically, when the gains of the SD- and SR-links are close, sub-scheme \(D^{(s)}\) (two-hop relaying) may not be applied at all. Indeed, in this case and even for low source rates, two-hop relaying under-uses the direct link to transmit information and is thus sub-optimal. Furthermore, given the power constraints, outage can occur before the sequences \((D^{(s)}, B^{(s)}, A^{(s)})\) and \((D^{(s)}, C^{(s)}, A^{(s)})\) are completed, such that \(A^{(s)}, B^{(s)}\) or \(C^{(s)}\) may not be applied. We next describe the condition of existence of these sub-schemes.

First, consider the sequence \((D^{(s)}, C^{(s)}, A^{(s)})\), i.e. two-hop relaying, full decode-forward without beamforming and partial decode-forward with beamforming. When the SR-link is the limiting one, S-EE is composed of only \((D^{(s)}, C^{(s)})\) for most channel realizations. Partial decode-forward with beamforming \((A^{(s)})\) is only applied if the relay power \(P_r\) is very low compared to the source power \(P_s\).

Second, we consider the sequence \((D^{(s)}, B^{(s)}, A^{(s)})\), i.e. two-hop relaying, full decode-forward with beamforming and partial decode-forward with beamforming. When the network is severely limited by both the SD- and RD-links, S-EE is composed of only \((D^{(s)}, B^{(s)})\). Recall that in \(B^{(s)}\), the relay constraint is already met. Thus, if the direct link is weak given the source power constraint, the scheme rapidly goes into outage and cannot achieve high rates.

Finally, S-EE can also be reduced to only two-hop relaying \((D^{(s)})\) when the direct link is almost null.

**D. Maximum rates achieved by the three optimized schemes**

We now analyze the maximum rate achievable for the three optimized schemes. Formal proof can be found in Appendix [12].

First, in N-EE (which minimizes the network consumption), the source power constraint is always met with equality before the relay power constraint. Therefore, at the expense of consuming more total energy, S-EE and R-EE (which respectively minimize the source and the relay consumption) allow the source to reach higher rates than the maximal rate achieved by scheme N-EE, denoted as \(R_{\text{max}}\).

Second, analyzing the outage S-EE sub-schemes gives a comprehensive description of the overall maximum achievable rate \(R_{\text{max}}\). Outage is declared when the power constraints cannot be met given the source rate, which can occur with any sub-scheme, as explained in Section [14]. For sub-schemes \(A^{(s)}\) and \(B^{(s)}\), the relay power constraint is already met and S-EE goes in outage as soon as the source power constraint is met. Therefore, in this case, at \(R = R_{\text{max}}\), both source and relay power constraints are met. On the contrary, when sub-schemes \(C^{(s)}\) and \(D^{(s)}\) are applied, the relay power constraint is not limiting. Thus, S-EE goes in outage only if the source power constraint is met. In this case, the maximum achievable rate \(R_{\text{max}}\) can be derived in closed-form. The values of \(R_{\text{max}}\) are described in Table [12]. Figure [6] gives an example of \(R_{\text{max}}\).
A power allocation for minimizing the relay energy consumption source rate is infeasible for N-EE, the message splitting and in (4) with equalized source power constraint (2). Once the R_B long as feasible, given the rate and power constraints. In for minimizing the total energy consumption are applied as denoted as G-EE, the message splitting and power allocation are always met with equality in both problems and, with rate and two power constraints. The two rate constraints are always met with equality in both problems and, with source rate up to R_{max}, G-EE is equivalent to the rate-optimal scheme. When R \rightarrow R_{max}, the source power constraint is also met with equality (G-EE applies R-E). Hence, in the interior neighbourhood of R_{max}, each optimization problem has just one degree of freedom left. When R \geq R_{max}, an outage occurs; either the desired source rate becomes infeasible, or the relay power constraint is also met with equality. Thus, above R_{max}, there is no more degree of freedom for neither problem and R_{max} is the maximal rate that is achieved for both G-EE and maximum-rate schemes.

Proof: If a source rate achieved with G-EE is not rate-optimal for this energy, then there exists another set of power allocation that consumes the same energy but achieves a better rate R_m. For this higher rate R_m, G-EE consumes more energy that the maximum-rate scheme and is thus suboptimal for energy, which is contradictory. Thus, for all achievable rates, G-EE is also rate-optimal. Also G-EE achieves all the rates that are achievable by the maximum-rate scheme (see below).

Considering energy as a function of rates and given fixed individual power constraints, the maximum source rate R_{max} achieved with energy minimization G-EE is equal to the maximum rate achieved by rate maximization.

Proof: Note that both G-EE and the maximum-rate scheme are optimization problems with four constraints (two rate and two power constraints). The two rate constraints are always met with equality in both problems and, with source rate up to R_{max}, G-EE is equivalent to the rate-optimal scheme. When R \rightarrow R_{max}, the source power constraint is also met with equality (G-EE applies R-E). Hence, in the interior neighbourhood of R_{max}, each optimization problem has just one degree of freedom left. When R \geq R_{max}, an outage occurs; either the desired source rate becomes infeasible, or the relay power constraint is also met with equality. Thus, above R_{max}, there is no more degree of freedom for neither problem and R_{max} is the maximal rate that is achieved for both G-EE and maximum-rate schemes.

Finally, both previous points imply that only the combined scheme G-EE is equivalent to the maximum-rate scheme. Figure 6 illustrates this equivalence. As a reference, we consider the rate-optimal scheme proposed in [6, 7]. This scheme is similar to A^{(n)} in coding: the source performs message splitting, sending one part during phase 1, and both during phase 2. Then, the power allocation is numerically optimized to maximize the achievable rate. For the comparison, we compute the maximum capacity achievable with the rate-optimal scheme based on the energy consumed using G-EE (plot with reversed axis).

C. Discussion

Even though both analysis and simulation show that the generalized scheme G-EE and the maximum-rate scheme proposed in [6, 7] lead to the same set of power allocation, we argue that the proposed scheme G-EE brings two significant advantages.

First, existing maximum-rate schemes proposed in the literature require numerical search of the optimal solution for each channel realization. On the contrary, the G-EE scheme,
which is both rate- and energy-optimal, has the optimal power allocation in closed-form.

Second, this energy efficiency approach reveals more details about sub-schemes and power allocations than the maximum-rate approach. We show that minimizing the network energy consumption is not equivalent to maximizing the network capacity as often believed, since the whole range of source rates achievable with decode-forward cannot be covered. Furthermore, we highlight that several regimes of coding technique and power allocation exist (partial or full decode-forward, with or without beamforming) and each of them corresponds to a specific energy optimization problem (N-EE or R-EE).

V. SIMULATION RESULTS

A. Description of the simulation environment

We simulate the performance of the proposed schemes in a realistic environment and consider both pathloss and shadowing. Shadowing is the most severe random attenuation factor encountered in urban cellular environment \[17\]. It refers to mid-term channel variations which are due to the multipath propagation generated by obstructions between the transmitter and receiver. Since blocking objects do not vary rapidly in geographical position, size or dielectric properties, shadowing remains slow and predictable. Therefore, in our case, we can reasonably assume that channel gains are constant over the two transmission phases and that nodes can track those variations at each instant.

We consider the situation depicted in Figure 5. In a dense urban environment with high surrounding buildings, a mobile user (source S) sends data to a base station (destination D) with the help of a fixed relay station (relay R), such as in LTE or Wimax. The mobile user is located outdoors, at street level. The relay station is below the top of surrounding building and the base station is far above the rooftop. This channel model corresponds to the Stationary Feeder scenario described in the European WINNER project \[18\]. Pathloss and shadowing effects are given by (4.23) of \[18\].

\[
PL = A \log_{10}(d[m]) + B + C \log_{10}\left(\frac{f_c[GHz]}{5.0}\right) + \chi
\]

where \(d\) is the distance between the transmitter and receiver and \(f_c\) is the frequency carrier (within the range 2-6GHz).

\(\chi\) refers to shadowing effect, it is modelled as a zero-mean log-normal random process, with standard deviation \(\sigma^2\). The parameters \(A, B, C\) and \(\sigma^2\) depend on the global location of the transmitters and receivers (street level, rooftop...). Considering the relative positions of nodes and using notations of \[18\], the SR-link is modelled by scenario B1, the SD-link by scenario C2 and the RD-link by scenario B5f. Table II gives the distances transmitter-receiver, as well as the pathloss and shadowing parameters for each link. For simulations, we take \(f_c = 3.5\)GHz, \(P_s = 100\)mW, \(P_r = 200\)mW, which refers to a Wimax situation. We also consider \(\theta = 0.3\) and \(N = 10^{-10}\)W.

B. Reference Schemes

As reference schemes, we consider direct transmission, two-hop transmission and the maximum-rate scheme in \[6\] \[7\]. Direct transmission is one-phase communications. The source sends the message \(m\) with no splitting (\(m_d = m\) and \(m_r = 0\)) and minimum power

\[
\eta_D P_s = \frac{(2^{R - 1}) N}{|h_d|^2}
\]

In the classical two-hop routing, the source first sends the whole message to the relay, again without splitting, and with a minimum power \(\rho_{2H}^s P_s\). If the relay can decode the message, it forwards to the destination with a power of \(\rho_{2H}^r P_r\), where

\[
\rho_{2H}^s P_s = \frac{(2^{R - \theta - 1}) N}{|h_s|^2}; \quad \rho_{2H}^r P_r = \frac{(2^{R - \theta - 1}) N}{|h_r|^2}
\]

The destination then decodes using only the signal received from the relay (not both signals received from source and relay as in the two-hop sub-scheme \(D^{(3)}\)).

The maximum-rate scheme in \[6\] \[7\] uses partial decode-forward with power allocation optimal for rate. Allocation is

\begin{table}[h]
\centering
\begin{tabular}{|c|c|c|c|c|}
\hline
Link & d[m] & A & B & C & \(\sigma^2\) \\
\hline
SR & 70 & 22.7 & 41.0 & 20 & 3 \\
SD & 100 & 26 & 39 & 20 & 4 \\
RD & 30 & 23.5 & 57.5 & 23 & 8 \\
\hline
\end{tabular}
\caption{PATHLOSS AND SHADOWING PARAMETERS}
\end{table}

Fig. 5. Communication in a realistic environment

Fig. 6. Total power consumption as function of the rate during both phases.
computed numerically solving a convex optimization problem. Whereas classical performance simulations plot capacity as a function of energy, we reverse the axes to plot energy as a function of rate and to allow comparison. We then analyze the performance of the proposed schemes in both non-fading and fading environments.

C. Performance in non fading environment

We first consider a non fading environment, with only pathloss ($\chi$ is set to zero). This can also be viewed as instantaneous performance.

Figures 6 and 7 respectively show the total power and the per-node power needed to maintain a range of desired source rates. When a source rate is not achievable, an outage occurs, which is depicted by a cut-off in the curve. These figures illustrate the analysis of Section III-D. First, higher rates are achieved by optimizing only one node consumption (either the source as in S-EE, or the relay as in R-EE), rather than the whole network consumption (N-EE). Moreover, both schemes S-EE and R-EE achieve the same maximum source rate. Second, the combined scheme G-EE is also rate-optimal and significantly outperforms both direct and two-hop routing for every achievable source rate. For example, at a rate of 1.35 normalized b/s/Hz, 1.2dB of energy gain is obtained over direct transmissions and 4.5dB over two-hop routing. The gain comes mainly from lessening the relay power consumption. Note that at $R = R_{\text{max}}$, the relay power constraint is not necessarily met with equality if rates beyond $R_{\text{max}}$ are infeasible due to rate constraints. Similarly, R-EE and S-EE provide significant power gain at the relay and the source respectively. Note that maximum-rate scheme is neither source nor relay energy-efficient as it consumes more power at each respective node than the optimal scheme for that node.

Two-hop routing generally suffers from very high instantaneous power consumption. Since the whole message is sent twice, one by the source and one by the relay, within the same time slot, both transmitters have to increase their transmit power to sustain the desired rate. We plot in Figure 8 the power consumed at each phase for the proposed and reference schemes. We see that the proposed schemes allocate power in a smooth manner and reduce transmit power peaks. Contrary to routing, energy is more evenly spread over both phases and over both direct and relaying paths. Moreover, the consumption per phase as function of the source rate is similar to the direct transmission consumption. This can help
managing interference in networks which allow both direct and relayed transmissions.

D. Performance in fading environment

We consider in this section both pathloss and shadowing. Due to random shadowing, there exists a non-zero probability that the channel gains are too weak to perform transmission. We plot in Figure 9 the average total network energy consumption. The cut-off in the curve depicts the maximum rate above which the outage probability is greater than 0.05. Simulations show that the proposed schemes significantly reduce the average energy consumed, as well as increase the maximal achievable source rate, given the outage requirement.

Finally, we analyze the utilization region of the proposed schemes and evaluate the energy gain as a function of the mobile user location, similarly to [19]. We consider the environment and heights of relay and base station as described in Section V-A. The mobile user is free to locate anywhere at street level. The relay station is located at the origin (0,0) and the destination at (20√2, 0) so that dRD = 30m. As reference, we consider a combined scheme which uses either direct transmissions (DTx) or two-hop routing (RTx), depending on the total consumed energy. For simulations, equal time division is assumed, with a source rate of 1bit/s/Hz. Figures 10 and 11 respectively depict the power gain in dB realized by the proposed generalized scheme G-EE over the reference combined scheme and the outage gain (computed as the difference between outage probabilities of reference and G-EE). Simulation shows that G-EE provides up to 2.2dB of energy gain can be obtained, as well as a reduction of 37.5% of the outage probability.

VI. CONCLUSION

We explore the issue of energy efficiency in relay channels and design a half-duplex coding scheme with optimal resource allocations that maintain a desired source rate and minimize the energy consumption. We consider the network, the relay and the source consumption separately and then combine them into a generalized set of power allocation G-EE which is energy optimal for the relay channel.

We show that minimizing the network energy consumption is not equivalent to maximizing the network capacity as often believed, since the whole range of source rates achievable with Decode-Forward cannot be covered by individual schemes. This equivalence only holds for the proposed generalized scheme G-EE.

Furthermore, this energy efficiency approach remains meaningful since it allows a comprehensive description of the optimal power allocation (full or partial decode-forward, with or without beamforming) as well as closed-form solutions of the optimal rate allocation, which remains implicit in maximum-rate schemes. Reversely, we highlight that rate-optimal scheme is not energy-efficient for either node.

The generalized scheme G-EE gives a practical algorithm for message coding and associated power allocation to reach energy efficiency and rate optimality for each desired source rate and channel realization. Applied in cellular model, G-EE is particularly beneficial at cell-edge where up to 2.2dB energy gain can be obtained, as well as a reduction of 37.5% of the outage probability.

APPENDIX A

PROOF OF ALGORITHM I: POWER ALLOCATION FOR THE NETWORK ENERGY EFFICIENCY

A. General problem setting

Denote X = (η1, ρ1, η2, ρ2, ρr)T . Consider optimization problem [5]. We look for the X that minimizes the objective function f, such that

\[ f(X) = \omega_s \left[ \theta (\eta_1 + \rho_1) P_s + \bar{\theta} (\eta_2 + \rho_2) P_s \right] + \omega_r \bar{\theta} \rho_r P_r \]

given the two rate constraints c1(X) = I1 - R and c2(X) = I2 - R, as defined in [4]. Here, we first ignore the power constraints. The Lagrangian can then be formed as

\[ \mathcal{L} = f(X) - \lambda_1 c_1(X) - \lambda_2 c_2(X) \]

Assuming natural logarithms does not change the optimal solution. The KKT conditions are expressed by equations (14) - (18) at the top of next page and by (19).

\[ \lambda_1 c_1(X) = 0 \quad c_1(X) \geq 0 \quad \lambda_1 \geq 0 \quad i = 1, 2 \]

B. Transmission during the first phase

Note that \( \frac{\partial \mathcal{L}}{\partial \eta_i} \) and \( \frac{\partial \mathcal{L}}{\partial \rho_i} \) cannot be equal to zero simultaneously, unless \( |h_d|^2 = |h_s|^2 \). Therefore, we have to relax either
\[
\frac{\partial \mathcal{L}}{\partial \eta_1} = \omega_s \theta P_s - \frac{\lambda_1 \theta \frac{P_s |h_d|^2}{\mathcal{N}}} \left[ 1 + \frac{(1 + \rho_1) P_s |h_a|^2}{\mathcal{N}} \right] - \frac{\lambda_2 \theta P_s}{\mathcal{N}} \left[ 1 + \frac{|h_a|^2}{\mathcal{N}} \right] - \frac{|h_d|^2}{\mathcal{N}} \left[ 1 + \frac{|h_a|^2}{\mathcal{N}} \right] = 0
\]

\[
\frac{\partial \mathcal{L}}{\partial \rho_1} = \omega_s \theta P_s - \lambda_1 \theta \frac{P_s |h_a|^2}{\mathcal{N}} \left[ 1 + \frac{(1 + \rho_1) P_s |h_a|^2}{\mathcal{N}} \right] - \frac{\lambda_2 \theta P_s}{\mathcal{N}} \left[ 1 + \frac{|h_a|^2}{\mathcal{N}} \right] = 0
\]

\[
\frac{\partial \mathcal{L}}{\partial \eta_2} = \omega_s \theta P_s - \lambda_1 \theta \left[ 1 + \frac{(\eta_2 + \rho_2) P_s |h_a|^2}{\mathcal{N}} \right] - \frac{\lambda_2 \theta P_s}{\mathcal{N}} \left[ 1 + \frac{|h_a|^2}{\mathcal{N}} \right] = 0
\]

\[
\frac{\partial \mathcal{L}}{\partial \rho_2} = \omega_s \theta P_s - \lambda_1 \theta \left[ 1 + \frac{\eta_2 P_s |h_a|^2}{\mathcal{N}} \right] - \frac{\lambda_2 \theta P_s}{\mathcal{N}} \left[ 1 + \frac{|h_a|^2}{\mathcal{N}} \right] = 0
\]

\[
\frac{\partial \mathcal{L}}{\partial \rho_r} = \omega_s \theta P_s - \lambda_1 \theta \left[ 1 + \frac{\eta_2 P_s |h_a|^2}{\mathcal{N}} \right] - \frac{\lambda_2 \theta P_s}{\mathcal{N}} \left[ 1 + \frac{|h_a|^2}{\mathcal{N}} \right] = 0
\]

From (16), \( \lambda_2 = \left( 1 - \frac{1}{\rho_1} \right) \left( \frac{N}{|h_d|^2} + \eta_2 P_s \right) > 0. \)

Thus, the second constraint is also active and we get

\[ \eta_2 P_s |h_d|^2 = G_s - 1 \quad (23) \]

From (20), (22) and (23), we write \( \rho_2, \rho_r, \) and \( \eta_2 \) as functions of \( \rho_1. \) Replacing the Lagrangian multipliers \( \lambda_1 \) and \( \lambda_2 \) in (15), we prove that \( \rho_1 \) solves \( g_1(\rho_1) = 0, \) where \( g_1 \) is defined by (7) of Algorithm I. Note that \( g_1 \) is decreasing in \( \rho_1, \) increasing in \( R \) and that \( g_1(0) = |h_a|^2 |h_d|^2 + |h_a|^2 |h_d|^2 2^{R/\theta} - 1 > 0 \) for all \( R. \) If, for all \( \rho_1 \in [0, 1/\theta], \) \( g_1(\rho_1) > 0, \) then the desired source rate is infeasible with the allocation as defined for \( A^{(n)} \) and subsequence \( B^{(n)} \) should be considered instead. We can also check that \( \rho_r > 0 \) and \( \rho_2 > 0 \) since \( |h_a|^2 > |h_d|^2 \) by assumption, and \( \eta_2 > 0 \) as long as \( R \geq \theta \log_2 \left( 1 + \frac{P_s |h_a|^2}{\mathcal{N}} \right) = R^{(n)} \)

On the contrary, when \( R \leq R^{(n)} \), \( \eta_2 \) is negative and there is no solution for (23). Thus, we relax \( \eta_2 \) and rewrite the optimization. Following the same steps as above, we can show that constraints are still active, and that (15), (17) and (18) still hold. This leads to Proposition 2. If the power constraints are not satisfied, then the desired source rate is infeasible.

**APPENDIX B**

**PROOF OF ALGORITHM [I]: POWER ALLOCATION FOR RELAY ENERGY EFFICIENCY**

The proof of Algorithm I and Propositions [I] and [II] follows the same main steps as in Appendix A. Here, \( \omega_s = 0 \) and \( \omega_r = 1. \) Analysis in Appendices A-A and A-B still holds. To minimize the energy consumption of the relay, direct transmissions should be performed as long as the direct link is not in outage, or equivalently as long as \( R \leq \log_2 \left( 1 + \frac{P_s |h_a|^2}{\mathcal{N}} \right) \)

(Proposition [II].)

The relay is needed to perform the transmission as soon as the source meets its power constraint and cannot send more.
data without going in outage. Then,
\[ \rho_2 = \frac{1 - \rho_1 \theta}{\theta} \] (24)

First, note that Eq. \( \frac{\partial L}{\partial \rho_2} = 0 \) [18] is still valid. Thus \( \lambda_1 = \Gamma_r \left[ \frac{|h_s|^2}{N \ln 2} \left( 1 + \sqrt{\frac{P_s|h_s|^2}{P_s|h_s|^2 + \rho}} \right) \right]^{-1} \). Second, with (24), the Lagrangian equations now satisfy
\[
\frac{\partial L}{\partial \eta_2} = \lambda_1 \frac{\partial P_s|h_s|^2}{N \ln 2} - \lambda_2 \frac{P_s|h_s|^2}{N \ln 2 \Gamma_r} + \lambda_2 \frac{P_s|h_s|^2}{N \ln 2 \Gamma_r} = 0
\] (25)
\[
\frac{\partial L}{\partial \rho_2} = \lambda_1 \frac{\partial P_s|h_s|^2}{N \ln 2} - \lambda_2 \frac{P_s|h_s|^2}{N \ln 2 \Gamma_r} + \lambda_2 \frac{P_s|h_s|^2}{N \ln 2 \Gamma_r} = 0
\] (26)

Subtracting (25) from (26), we get
\[
\lambda_2 = \lambda_1 \frac{P_s|h_s|^2}{P_s|h_s|^2} \left( 1 + \frac{\eta_2 P_s|h_s|^2}{N} \right) \Gamma_r^{-1

Both constraints are active: Eq. (21) \( G_d = \Gamma_r \) and Eq. (22) \( \eta_2 P_s|h_s|^2 = G_s - 1 \) still hold. From (23), \( \eta_2 \) is expressed as a function of \( \rho_1 \). From (21), (24) and the expression of \( \eta_2, \rho_c \) can be also written as a function of \( \rho_1 \).

Finally, plugging the expressions of \( \lambda_1 \) and \( \lambda_2 \) in (21), (23) into (26), we prove that \( \rho_1 \) should satisfy \( g_2(\rho_1) = 0 \), as defined by 9 of Algorithm II with
\[
g_2(\rho_1) = \sqrt{\frac{P_s|h_s|^2 P}{P_s|h_s|^2}} \frac{2^{R/\theta} - \rho_1 \frac{P_s|h_s|^2}{N}}{1 + \rho_1 \frac{P_s|h_s|^2}{N}}^\theta /
= \sqrt{\frac{P_s|h_s|^2 P}{P_s|h_s|^2}} \frac{2^{R/\theta} - \rho_1 \frac{P_s|h_s|^2}{N}}{1 + \rho_1 \frac{P_s|h_s|^2}{N}}^\theta /
\]

Now, we focus on the existence of this power allocation set. First, we note \( \eta_2 > 0 \). \( R \) must be greater than \( \log_2 \left( 1 + \frac{\rho_1 P_s|h_s|^2}{N} \right) = R_2^{(r)} \). Moreover, \( \rho_r \) is positive as long as \( R > \log_2 \left( 1 + \frac{\rho_1 P_s|h_s|^2}{N} \right) + \theta \log_2 \left( 1 + \frac{\eta_2 + \rho_1 P_s|h_s|^2}{P_s|h_s|^2} \right) = \log_2 \left( 1 + \frac{P_s|h_s|^2}{N} \right) = R_1^{(r)} \). Finally, the existence of \( \rho_1 \), and consequently of \( \rho_2 \), follows the same steps as in Appendix A. This concludes the proof of Proposition 5.

When \( R < \log_2 \left( 1 + \frac{\rho_1 P_s|h_s|^2}{N} \right) \), \( \eta_2 \) is negative and there is no solution for (23). In this case, we relax \( \eta_2 \) and rewrite the optimization problem. As above, we can show that rate constraints are still active, and that (18) and (26) still hold. This concludes the proof of Proposition 4 and Algorithm II.

APPENDIX C

PROOF OF ALGORITHM III: POWER ALLOCATION FOR SOURCE ENERGY EFFICIENCY

The proof of Algorithm III and Propositions 7, 8 and 9 follows the same steps as in Appendix A. Here, \( \omega_r = 1 \) and \( \omega_1 = 0 \). Analysis of Lagrangian in A-A and A-B still holds. However, the constraint on the relay consumption is relaxed here, such that it can consume up to its maximal power.

First, let’s consider that the source rate is high and that all the available relay power is required (\( \rho_r = \frac{1}{\theta} \)). This means that the first rate constraint, which depends on \( \rho_r \), will necessarily be tight. From Eq. \( \frac{\partial L}{\partial \rho_r} = 0 \) [15] and Eq. \( \frac{\partial L}{\partial \rho_2} = 0 \) [17], we can prove that both Lagrangian multipliers are positive, meaning that both constraints are active, such that
\[
\lambda_1 = \frac{|h_d|^2}{N \ln 2} \left( 1 + \sqrt{\frac{P_s|h_s|^2}{P_s|h_s|^2 + \rho}} \right) = G_d \ln 2 \left( 1 + \sqrt{\frac{P_s|h_s|^2}{P_s|h_s|^2 + \rho}} \right)
\]
\[
\lambda_2 = \left( 1 - \lambda_1 \frac{|h_s|^2}{N \ln 2 \Gamma_d} \right) \frac{N \ln 2 \Gamma_d}{|h_s|^2}
\]

Plugging in these expressions of \( \lambda_1 \) and \( \lambda_2 \), as well as Eq. \( G_d = \Gamma_r \) [21] and Eq. \( \eta_2 P_s|h_s|^2 = G_s - 1 \) [25] into (17), we prove that \( \rho_1 \) should satisfy \( g_2(\rho_1) = 0 \), as defined by (12) of Algorithm III. The existence of this power allocation set follows the same steps as for Appendices A and B. This leads to sub-scheme \( A^{(s)} \) and Proposition 8. Note that the power allocation is similar to sub-scheme \( A^{(r)} \). We now have to insure the non-negativity of this solution, in particular \( \eta_2 \geq 0 \) and \( \rho_2 \geq 0 \).

On the one hand, we consider the non-negativity constraint on \( \eta_2 \). \( \eta_2 \) is positive following sub-scheme \( A^{(s)} \) as long as \( R \geq R_{2,C}^{(s)} \). If \( R \leq R_{2,B}^{(s)} \), \( \eta_2 \) is relaxed and set to 0. Next, the optimization follows the same steps as for \( B^{(n)} \) or \( B^{(r)} \), but with \( \rho_r = \frac{1}{\theta} \). This leads to sub-scheme \( B^{(s)} \) and Proposition 8. Note that, following \( B^{(s)} \), \( \rho_2 \) is positive as long as \( R \geq R_{1,B}^{(s)} \), with \( R_{1,B} = R_B \) as defined in Algorithm III. This means that the relay requires all its power to forward the source data such that the first rate constraint \( c_1 \) is active. However, if the rate decreases under the threshold \( R_1^{(s)} \), the relay may not need all its available power and the first rate constraint becomes inactive. \( \rho_2 \) is relaxed and set to 0. Finally, even if the relay consumption is not part of the optimization problem here, we can nevertheless reduce it to the minimum required, such that \( R = R_1^{(s)} \) (however, \( c_1 \) is still inactive in the sense of Lagrangian optimization). Thus, we deduce the expression of \( \rho_r < 1/\theta \). The optimization now leads to sub-scheme \( D^{(s)} \) and Proposition 8.

On the other hand and following sub-scheme \( A^{(s)} \) again, we consider the non-negativity constraint on \( \rho_2 \) (with \( \eta_2 > 0 \)). \( \rho_2 \) is positive following sub-scheme \( A^{(s)} \) as long as \( R \geq R_{2,C}^{(s)} \), with \( R_{2,C}^{(s)} \) = \( \theta \log_2 \left( \frac{P_s|h_s|^2}{\eta_2 N} \right) - \frac{1}{\eta_2 N} \left( 1 + \frac{\theta P_s|h_s|^2}{N} \right)^{1/\theta} \).
If $R \leq R_{t1,C}^2$, $\rho_2$ is relaxed and set to 0. This also means that the relay is able to forward the source data without requiring the beamforming gain between the source and the relay. Therefore, the first rate constraint $c_1$ becomes inactive and $\lambda_1 = 0$. From Eq. $\frac{\partial C}{\partial \eta_2} = 0$ [16] and Eq. $\frac{\partial C}{\partial \eta_2} = 0$ [15], we get $\frac{|a_1|^2}{1 + 2|a_1|^2} = c_2 |a_1|^2$, such that $\eta_2 = \rho_1 + \frac{|a_1|^2}{N \rho_1 |a_1|^2}$. As the second constraint $c_2$ is active, Eq. $\frac{\eta_2 P_s |a_1|^2}{1 + 2|a_1|^2} = G_s - 1$ [23] is still valid and we get $\eta_2 = \frac{N}{P_s |a_1|^2} \left( \frac{2N}{1 + \frac{2|a_1|^2}{\rho_1}} - 1 \right)$. By equalizing both expressions of $\eta_2$, we deduce $\rho_1$. Next, even if the relay consumption is not considered here, we reduce it to its minimum required value. The optimization leads to sub-scheme $C(s)$ and Proposition 7. Now, following $C(s)$, $\eta_2$ is positive as long as $R \geq R_{t1,C}^2$, with $R_{t1,C} = R_C$ as defined in Algorithm III. If the rate decreases under the threshold $R_{t1,C}$, $\eta_2$ is relaxed and set to 0, which leads to $D(s)$ as previously defined.

So far, depending on non-negativity constraints, we defined two possible sequences of sub-schemes: $(A(s), B(s), D(s))$ and $(A(s), C(s), D(s))$. We now define how each of them should be used. First, $A(s)$ is applied when $R > R_{t1,C}$. If $R_{t1,C}^2 \leq R_{t2,C}$, the non-negativity constraint on $\eta_2$ is stronger than the constraint on $\rho_2$ and the sequence $(A(s), B(s))$ is applied. Otherwise, $(A(s), C(s))$ is applied. Second, $D(s)$ is such that $\eta_2 = 0$ and $\rho_2 = 0$. This sub-scheme is applied as long as $R \leq \min \left\{ R_{t1,B}^2, R_{t1,C}^2 \right\} = R_1(s)$. Thus, if $R_{t1,B}^2 \leq R_{t1,C}$, the sequence $(B(s), D(s))$ is applied, otherwise $(C(s), D(s))$ is applied.

Finally, we show that upper and lower rate bounds for applying either $B(s)$ or $C(s)$ are consistent. Considering the source consumption as a function of the source rate, sub-schemes are continuous. Moreover, $\eta_2$ and $\rho_2$ are increasing functions of the source rate. Thus, if $R_{t1,B}^2 \leq R_{t1,C}$ and sequence $(D(s), B(s))$ is applied for low source rates, we get $\rho_2 > 0$ for any $R > R_{t1,B}$. In this case, we necessarily have $R > R_{t2,B}$, which implies $R_{t2,B}^2 \leq R_{t2,B}$. Therefore, only sequence $(A(s), B(s), D(s))$ can occur. Similarly, we can show that only $(A(s), C(s), D(s))$ can occur if $R_{t1,B}^2 \geq R_{t1,C}$. This completes the proof of Algorithm III.

**APPENDIX D**

**MAXIMAL ACHIEVABLE SOURCE RATE**

**A. R-EE and S-EE achieve higher rates than N-EE**

A source rate $R$ is achievable for N-EE (resp. R-EE) as long as there exists a $\rho_1^s$ satisfying $g_1(\rho_1^s) = 0$ (resp. $g_2(\rho_1^s) = 0$). Recall that $g_1(0) > 0$, $g_2(0) > 0$ and that both functions are decreasing (see Appendix A). Since $g_2(x) < g_1(x)$, there exists a range of $\rho_1$ for which $g_1(\rho_1) > 0$ but $g_2(\rho_1) \leq 0$. Therefore, for this range, we can find a solution for $g_2(\rho_1) = 0$ but not for $g_1(\rho_1^s) = 0$. The corresponding range of source rates is thus achieved with R-EE, but not with N-EE. The proof is similar for S-EE.

**B. Derivation of $R_{\text{max}}$ with sub-schemes $D(s)$ and $C(s)$**

If sub-scheme $D(s)$ is applied when S-EE goes in outage, the source power constraint is met, such that $(\rho_1^s + \eta_2^s) \theta P_s + (\rho_2^2 + \eta_2^2) \theta P_s = P_s$. This leads to $\rho_1^s = \frac{1}{\theta}$ and $R_{\text{max}} = \theta \log_2 \left( 1 + \frac{\rho_1^s |a_1|^2}{\theta N} \right)$.

Similarly, if sub-scheme $C(s)$ is applied when S-EE goes in outage, $(\rho_1^s + \eta_2^s) \theta P_s + (\rho_2^2 + \eta_2^2) \theta P_s = P_s$. Thus, $\theta \rho_1^s + \theta \left( \rho_1^s + \frac{\eta_2}{P_s} \left( \frac{|a_1|^2}{1 + 2|a_1|^2} \right) - \frac{|a_1|^2}{P_s} \right) = 1$ and $\rho_1^s = 1 + \frac{\eta_2 \theta}{P_s} \left( \frac{|a_1|^2}{1 + 2|a_1|^2} - 1 \right) = \frac{\eta_2 \theta}{P_s} \left( 2R \left( \frac{|a_1|^2}{|a_1|^2} \right) - 1 \right)$, from which we deduce $R_{\text{max}}$ as in Section II-D.

**C. S-EE and R-EE achieve the same maximum rate $R_{\text{max}}$**

We want to prove that the power allocation of S-EE for $R = R_{\text{max}}$ corresponds to the allocation of R-EE at this specific rate such that both achieve the same maximum rate $R_{\text{max}}$. Also note that as long as $|a_1|^2 > |h_1|^2$, the maximum rate achieved by R-EE using $C(r)$ (direct transmission) is strictly lower to the maximum rate achieved by R-EE using $B(r)$ or $A(r)$. Thus, when R-EE declares outage, either sub-scheme $B(r)$ or $A(r)$ is applied. Now, we successively consider the four possible cases for S-EE.

First, assume that $A(s)$ is applied when S-EE goes in outage at $R = R_{\text{max}}$. This means that $R \leq \theta \log_2 \left( 1 + \frac{\rho_1^s |a_1|^2}{\theta N} \right)$ where $\rho_1$ solves $g_2(\rho_1) = 0$. Therefore, R-EE uses $A(r)$ at this rate and $\rho_1 = \rho_1^s$. Now, note that $\eta_2^r = \eta_2^s$, since both solve the same expression. For R-EE, $\rho_1^r = 1 - \frac{\eta_2^r}{\theta}$ and S-EE declares outage when the source power constraint is met. Thus, $\rho_1^r = \rho_1^s$ at $R = R_{\text{max}}$. Noting that $\rho_2$ and $\rho_1$ satisfy the same equation in both R-EE and S-EE ($\Gamma_r = G_d$), we get $\rho_1^r = \rho_1^s = 1 - \frac{1}{\theta}$, which means that the relay constraint is also achieved and R-EE goes in outage for this same rate.

Second, assume that $B(s)$ is used when S-EE goes in outage. This means that $R \leq \theta \log_2 \left( 1 + \frac{\rho_1^s |a_1|^2}{\theta N} \right)$ where $\rho_1$ solves $g_2(\rho_1) = 0$ and that R-EE uses $B(r)$ when outage occurs. Similarly to the above analysis, we can show that the power allocation of S-EE for $R = R_{\text{max}}$ corresponds to the allocation of R-EE at this specific rate.

Third, we assume that $C(s)$ is used when S-EE goes in outage. In this case, we have, by concavity,

$$R_{\text{max}} = \theta \log_2 \left( 1 + \frac{\rho_1^s |a_1|^2}{\theta N} \right) + \frac{|a_1|^2}{|a_1|^2} - \theta \log_2 \left( \frac{|a_1|^2}{|a_1|^2} \right) \geq \theta \log_2 \left( 1 + \frac{\rho_1^s |a_1|^2}{\theta N} \right) + \frac{|a_1|^2}{|a_1|^2} - \theta \log_2 \left( \frac{|a_1|^2}{|a_1|^2} \right) = \theta \log_2 \left( 1 + \frac{\rho_1^s |a_1|^2}{\theta N} \right).$$

Thus, for all $\rho_1 \leq \frac{1}{\theta}$ solving $g_2(\rho_1) = 0$, we get $R_{\text{max}} \geq \theta \log_2 \left( 1 + \frac{\rho_1^s |a_1|^2}{\theta N} \right)$ such that at $R = R_{\text{max}}$. R-EE uses $A(r)$. Note that S-EE declares outage using $C(s)$ if the source rate is met, such that $\eta_2^r = \frac{1 - \rho_1^r}{\theta}$. Thus, we can show that the allocation of S-EE is also a solution for $A(r)$ at this specific source rate. Similar analysis can be done for $D(s)$. 


Therefore, for all cases, the power allocation of S-EE for $R = R_{\text{max}}$ corresponds to the allocation of R-EE at this specific rate and both S-EE and R-EE achieve the same maximum rate $R_{\text{max}}$.

**APPENDIX E**

**ANALYSIS OF SCHEME G-EE: CONTINUITY AND DIFFERENTIABILITY**

First, we show the continuity of both schemes N-EE and R-EE at $R = R_{\text{max}}^{(n)}$. Note that $\eta_1$ and $\eta_2$ solve the same equations for both sub-schemes $A^{(n)}$ and $A^{(r)}$. Then, since the first rate constraint is active, we get $g_s(\rho_1^*) = \sqrt{P_s h_1^2 \rho_2}$. In $A^{(n)}$, $g_3(\rho_1^*) = \frac{|h_1|^2}{|h_2|^2}$, such that $\rho_1^*$ and $\rho_2^*$ of $A^{(n)}$ are also optimal for $A^{(r)}$. Finally, at $R = R_{\text{max}}$, the source power constraint is met, such that $\rho_2^* = \frac{|h_1|^2}{|h_2|^2} - \eta_2^*$ for both sub-schemes.

Therefore, the optimal power allocation set for $A^{(n)}$ is also optimal for $A^{(r)}$ at this particular rate and schemes N-EE and R-EE are continuous.

Second, note that parameters $\rho_1$, $\eta_2$, $\rho_2$ and $\rho_r$ are not differentiable themselves at $R = R_{\text{max}}$, but that the energy of G-EE (the weighted sum of $\rho_1$, $\eta_2$, $\rho_2$ and $\rho_r$) is differentiable. To prove it, let’s consider the left and right derivatives of G-EE. The left derivative (resp. the right derivative) corresponds to the derivative of the total energy consumed by N-EE (resp. R-EE) at $R = R_{\text{max}}^{(n)}$. If both semi-derivatives are equal, then G-EE is differentiable at this source rate. Let’s write both as functions of $\frac{\partial G}{\partial R}$ and $\frac{\partial G}{\partial R}$. Also note that $\frac{\partial G}{\partial R} = \frac{\ln N}{\theta} G_s - \theta P_s |h_1|^2 G_s |h_2|^2 G_s$.

Let’s consider the left derivative (scheme N-EE). In this case, we have:

$$\frac{\partial E_{N,EE}}{\partial R} = \frac{\partial E_{N,EE}}{\partial R} + \frac{\partial E_{N,EE}}{\partial R}$$

$$= \theta P_s \frac{\partial (G_r)}{\partial R} + \frac{\partial G_r}{\partial R}$$

$$= \theta P_s \frac{\partial (G_{r})}{\partial R} + \left( \frac{\partial h_r}{\partial N} \right)$$

$$= \theta P_s \frac{\partial (G_{r})}{\partial R} + \left( \frac{\partial h_r}{\partial N} \right)$$

Since the source power constraint is met with equality, the right derivative (scheme R-EE) is such that

$$\frac{\partial E_{R,EE}}{\partial R} = \theta P_s \frac{\partial (G_r)}{\partial R} + \frac{\partial G_r}{\partial R} = \frac{\partial G_r}{\partial R} - \frac{\partial G_r}{\partial R}$$

$$- \frac{\partial P_r}{\partial R}$$

Since $\frac{\partial \rho_r}{\partial N} = \frac{\partial \rho_r}{\partial N} - \frac{\partial \rho_r}{\partial N}$, then

$$\frac{\partial E_{R,EE}}{\partial R} = \theta P_s \frac{\partial (G_r)}{\partial R} + \frac{\partial G_r}{\partial R}$$

Now, recall that both N-EE and R-EE are continuous at this particular rate. Thus, $\rho_2 = \frac{P_s}{P_s h_2^2 \rho_1}$ and the right derivative of G-EE is also such that

$$\frac{\partial E_{R,EE}}{\partial R} = \theta P_s \frac{\partial (G_r)}{\partial R} + \frac{\partial G_r}{\partial R}$$

which is similar to the left derivative $\frac{\partial E_{N,EE}}{\partial R}$. Now, define the constant $C$ (that may depend on channel gains or on the source rate) such that $\frac{\partial E_{R,EE}}{\partial R} = \frac{\partial E_{N,EE}}{\partial R} + C$. We get $\frac{\partial E_{R,EE}}{\partial R} = \frac{\partial E_{N,EE}}{\partial R} - C \theta P_s |h_1|^2 G_s |h_2|^2 G_s$.

Since $g_1(\rho_1) = 0$, then $\frac{\partial E_{R,EE}}{\partial R} = \frac{\partial E_{N,EE}}{\partial R}$ at $R = R_{\text{max}}$, which completes the proof.

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