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Wireless Mesh Networks Allowing Intra-Link Errors: CEO Problem Viewpoint

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Abstract—In this paper, we re-formulate an issue related to wireless mesh networks (WMNs) from a Chief Executive Officer (CEO) problem viewpoint, and provide a practical solution to a simple case of the problem. The problem is described as follows: an originator broadcasts its binary source information to several forwarding nodes over Binary Symmetric Channels (BSC); the originator’s source information suffers from independent random binary errors; at the forwarding nodes, they just further interleave, encode the received bit sequence, and then forward it, without correcting errors that may be happening in the originator-forwarding node links (intra-links), to the final destination (FD) over Additive White Gaussian Noise (AWGN) channels. This strategy reduces the complexity of the forwarding node significantly. A joint decoding technique at the FD is proposed by using the correlation knowledge between intra-links. The bit-error-rate (BER) performances show that the originator’s information can be reconstructed at the FD even by using a very simple coding scheme. Moreover, we provide BER performance comparison between joint decoding and separate decoding strategies. The optimization of coding rate at originator side is also considered in this paper. The simulation results show that excellent performance can be achieved by the proposed system. Furthermore, extrinsic information transfer (EXIT) chart analysis is performed to investigate convergency property of the proposed technique.

I. INTRODUCTION

The research on cooperative communications using joint source, channel and network coding, has recently attracted a lot of attention with the recognition of significant importance. In cooperative communication systems, antennas of the mobile devices are shared by the multiple users, with the aim of configuring a virtual multi-terminal environment, and thereby advantageous points of network-level cooperation, rather than assembly of point-to-point (P2P), is expected to be significant. Wireless Mesh Networks (WMNs) are a form of cooperative communications system, in which multiple nodes cooperate to relay messages to the destinations. High data throughput, power and spectral efficiencies, as well as better resource utilization are expected. WMN systems usually consist of a group of fixed or mobile devices and hence can be deployed smoothly and flexibly in complicated environments such as in devastated and emergency situations, tunnels, oil rigs, and/or for battlefield surveillance.

A primary goal of this paper is have an insight of the WMN allowing intra-link errors from the viewpoint of the CEO problem. We investigate a simple system of WMN that two forwarding nodes assist to transmit the originator’s message to the final destination (FD), as exemplified in Fig. 1. We model this system by a distributed remote source coding model, which is called Chief Executive Officer (CEO) problem [1] in Network Information Theory.

The terminology CEO problem originates from the situation where a CEO aims to estimate the source information which cannot be observed directly; N agents, referred to as forwarding nodes, are deployed to observe/relay independently corrupted versions of source information sequence \( u \); the observations \( u_i \), \( i = 1, 2, \cdots, N \) are separately encoded and forwarded to the FD over rate-constrained channels; the FD, which works like a chief executive officer of an organization, tries to reconstruct the source information sequence from the noisy observations while keeping the distortion lower than an acceptable level. Fig. 2 shows an abstract scenario of the CEO problem. Successive coding strategy for Gaussian CEO problem is considered in [2], [3]. So far, the quadratic Gaussian CEO problem, where the source \( u \) is represented by a memoryless Gaussian codebook and suffers from independent and identically distributed (i.i.d) Gaussian noise \( W_{1\cdots N} \), has already been solved [4]. However, CEO problems in many other cases are still left as open question.

In this work, we assume that the source information is a binary bit sequence, and suffers from random binary errors.

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Furthermore the channels from the forwarding nodes to the FD suffers from Additive White Gaussian Noise (AWGN). Joint decoding is performed at the FD by utilizing the correlation knowledge between the forwarding nodes.

The rest of this paper is organized as follows. The system model used in this work is described in Section II. The proposed joint decoding strategy is detailed in Section III. In Section IV, we analyze the convergence property of proposed technique by using the extrinsic information transfer (EXIT) chart. Results of simulations conducted to verify the performance of the proposed technique are shown in Section V. A short discussion on rate optimization for the originator and forwarding nodes’ encoder is then provided in Section VI. Finally, the conclusions are drawn with some concluding remarks in Section VII.

II. SYSTEM MODEL

Fig. 3 shows a simple model of WMN, describing the viewpoint of the CEO problem, with only two forwarding nodes, assumed in this paper. The case of involving more forwarding nodes is further presented in [5]. For the encoders $C_0$, $C_1$ and $C_2$ in this model, we use only memory-1 $(2,3)_8$ half rate $(R = 1/2)$ convolutional codes. The structure of doped-accumulator (ACC) can be found in references [6]. As noted in [6], ACC is a rate-1 systematic recursive convolutional code with which every $P$-th systematic bits are replaced by the accumulated coded bits.

At the originator node, the original information bit sequence $x$ to be transmitted is first encoded by $C_0$. The encoded bit sequence is then interleaved by random interleaver $\Pi_0$ and doped-accumulated by ACC with doping ratio $P_{ori}$. The output of ACC, $u$, is broadcasted to the two forwarding nodes over independent Binary Symmetric Channels (BSC) with crossover probabilities $p_1$ and $p_2$, respectively, which can be modeled by the bit-flipping model [7].

Aiming at perfect decoding at the forwarding nodes is out of the scope of this paper, because it needs very strong link-level codes. Instead, each forwarding node makes only tentative decision on the received bit sequences, of which results $u_1$ and $u_2$ are first permuted by $\Pi_1$ and $\Pi_2$ and further encoded by $C_1$ and $C_2$, respectively. The channel between the transmitted information sequence $u$ and the one $u_i$ obtained as the result of the tentative decoding can be seen as also BSCs with the crossover probabilities $p_1$ and $p_2$, which can be modeled by the bit-flipping model, where $i = 1, 2$ is the node index.

Since perfect recovery of information sequence transmitted from the originator node is not aimed at in the forwarding nodes, the complexity of forwarding nodes is very light. Then, the encoded sequences are again interleaved by $\Pi_1$ and $\Pi_2$, and doped-accumulated by ACC with doping ratio $P_{for}$. The doped-accumulated bit sequences are modulated by BPSK, i.e., $0 \rightarrow -1$ and $1 \rightarrow +1$, and then forwarded to the FD at different time slots over AWGN channels. We assume the signal-to-noise ratios (SNRs) are the same in the two channels between forwarding nodes and the FD. The received signal sequences can be expressed as:

$$y_{i,j} = s_{i,j} + n_j,$$  \hspace{1cm} (1)

where $i = \{1, 2\}$ denotes the index of the forwarding nodes and $j$ is the symbol timing index. The modulated signal sequence is denoted by $s_{i,j}$; $n_j$ represents noise that are zero mean i.i.d complex value with variance $\sigma^2$ per dimension.

After receiving the signal $y_{i,j}$, the channel log-likelihood ratios (LLRs) are first calculated by:

$$L_{i,j}^{ch} = \log \frac{\Pr(y_{i,j}|s_{i,j} = +1)}{\Pr(y_{i,j}|s_{i,j} = -1)} = \frac{2}{\sigma^2} y_{i,j}. \hspace{1cm} (2)$$

At the FD, joint decoding is performed by exchanging the extrinsic LLRs which is detailed in section III.

III. JOINT DECODING STRATEGY

Joint decoding process is divided into two iteration processes as depicted in Fig. 3. We refer these two processes to Horizontal Iteration (HI) and Vertical Iteration (VI). To perform the channel decoding for convolutional codes $C_0$, $C_1$ and $C_2$ as well as for ACC, we perform Maximum A Posteriori (MAP) decoding using the BCJR algorithm proposed by Bahl, Cocke, Jelinek and Raviv [8].

In the HI, the extrinsic LLRs are exchanged through the corresponding interleaver/de-interleaver between the soft-in-soft-out (SISO) decoder $ACC^{-1}$ and SISO decoder $C_1^{-1}$ or $C_2^{-1}$ used by the forwarding node 1 and 2, respectively. The extrinsic information exchange is performed via HI until no more significant mutual information (MI) improvement can be achieved. However, activation times on the two HI loops are design parameters and hence optimization of the activation ordering is out of the scope of this paper. After each HI step, we activate VI loop between $C_1^{-1}$ and $C_2^{-1}$ by exchanging the output extrinsic LLRs of uncoded (systematic) bits output from the two decoders $C_1^{-1}$ and $C_2^{-1}$ via an LLR updating function $f_c$. The purpose of VI is to help two decoders $C_1^{-1}$ and $C_2^{-1}$ cooperate each other to reconstruct information. This is because since the uncoded bit sequences, are originated from the common originator node and forwarded by the two forwarding nodes, they are correlated. Hence, the aim is to fully exploit the knowledge about the correlation at the FD node.

After performing LLR exchange several times via the HI-VI loops, the a posteriori LLRs output from $C_1^{-1}$ and $C_2^{-1}$

![Fig. 2. An abstract model of CEO problem](image-url)

are combined. Before combing, however, the LLRs are further modified by (3) [9] [10]. We assume that the error probabilities values \( p_1 \) and \( p_2 \) are estimated at the forwarding nodes, and the estimates are forwarded to the FD with the help of higher layer protocols, however, they may be estimated at the FD only similarly to [6].

\[
L_{comb}^{i,j} = \sum_{i=1}^{2} L_{i,j}^m = \sum_{i=1}^{2} \log \frac{1 - p_i}{p_i} \text{sign}(L_{i,j}^p). \tag{3}
\]

where, \( p_i \) represents the error probabilities of the originator-forwarding node link referred to as intra-link. \( L_{i,j}^p \) are the a posteriori LLRs from \( C_{i}^{-1} \) and \( C_{2}^{-1} \). The function \( \text{sign}(\cdot) \) takes the sign (positive or negative) of its argument.

The modified LLRs \( L_{comb}^{i,j} \) are forwarded to another horizontal iteration loop to finally obtain the originator’s source information bit sequence \( \bar{x} \). This process is the same as the HI described above. Finally, hard decisions are made on the a posteriori LLRs obtained by the decoder \( C_{0}^{-1} \).

A. LLR updating function

As described above, the received bits sequences at two forwarding nodes are transmitted from the same originator, but corrupted by errors happening at different positions. The technique described above aims to utilize the correlation knowledge between them to achieve better performance. Although, \( p_i \) may be estimated from the a posteriori LLRs of the uncoded bits, we assume \( p_i \) are known at the FD that in this work.

It is very straightforward according to [7] that we can obtain the following equation:

\[
\Pr(u_1 = 0) = (1 - p_1) \Pr(u = 0) + p_1 \Pr(u = 1) \tag{4}
\]

\[
\Pr(u_1 = 1) = (1 - p_1) \Pr(u = 1) + p_1 \Pr(u = 0) \tag{5}
\]

Substituting (5) into (4), we can derive the probability updating function between the two forwarding nodes, as:

\[
\Pr(u_1 = 0) = (1 - \hat{p}) \Pr(u_2 = 0) + \hat{p} \Pr(u_2 = 1) \tag{6}
\]

\[
\Pr(u_1 = 1) = (1 - \hat{p}) \Pr(u_2 = 1) + \hat{p} \Pr(u_2 = 0)
\]

where \( \hat{p} = p_1 + p_2 - 2p_1p_2 \). Eq. (6) is equivalent to the LLR updating function \( f_c \), shown in (7), based on the value \( \hat{p} \) and the extrinsic LLRs of the uncoded bits obtained as the results of \( C_{1}^{-1} \) and \( C_{2}^{-1} \).

\[
L_{u_{1,j}}^a = f_c(L_{u_{2,j}}^e, \hat{p}) = \log \frac{(1 - \hat{p}) \cdot \exp(L_{u_{2,j}}^e) + \hat{p}}{(1 - \hat{p}) + \hat{p} \cdot \exp(L_{u_{2,j}}^e)}, \tag{7}
\]

where, \( L_{u_{2,j}}^e \) denotes the extrinsic LLRs of the uncoded bit sequence from \( C_{2}^{-1} \). By performing the operation \( f_c \), the extrinsic LLRs are updated, which takes into account the intra-link errors. The LLR updating function for \( u_2 \) can be easily obtained in the similar way.

IV. EXIT CHART ANALYSIS

Result of EXIT chart analysis [11] that indicate the convergence properties of the proposed system is provided in this section. The EXIT chart analysis covers the iterations \( ACC_{-1} \Leftarrow C_{i}^{-1} \) and \( ACC_{-1} \Leftarrow C_{0}^{-1} \), where notation ”\( \Leftarrow \)” denotes LLR exchange. We use three-dimensional (3D) EXIT chart to visualize the extrinsic information exchange between \( ACC_{-1} \) and \( C_{1}^{-1} \) as well as \( C_{1}^{-1} \) and \( C_{2}^{-1} \). As shown in the Fig. 3, two a priori LLR sequences are fed to \( C_{1}^{-1} \). Therefore, we evaluate the transfer function for \( C_{1}^{-1} \) in the form of:

\[
I_{c_{1}}^e = T_{1}(L_{c_{1},j}^e, f_c(L_{u_{2,j}}^e, \hat{p})), \tag{8}
\]

where \( I_{c_{1}}^e \) represents the mutual information between the LLRs of the coded bits obtained by \( C_{1}^{-1} \), and their corresponding transmitted bits . \( L_{u_{2,j}}^e \) denotes the a priori LLRs provided into \( C_{1}^{-1} \), which is equivalent to the extrinsic LLRs of coded bits output from \( ACC_{-1} \). The extrinsic LLRs of the uncoded bits output from \( C_{2}^{-1} \), \( L_{u_{2,j}}^e \), is updated by the \( f_c \) function and then fed to \( C_{1}^{-1} \). The transfer function for \( C_{2}^{-1} \) can be obtained in the same way as (8) was derived.

The 3D EXIT chart illustrated in Fig. 4 is with parameters BSC crossover probabilities \( p_1 = 0.05 \), \( p_2 = 0.06 \) and doping ratio \( P_{for} = 2 \) at SNR = -3.6 dB for both the channels between the forwarding nodes and the FD. Since the \( C_{1}^{-1} \) and \( C_{2}^{-1} \) are symmetric, we only show the 3D EXIT chart of the HI loop for the transmission chain of the forwarding node 1 and trajectory which is obtained by evaluating the mutual information between extrinsic LLRs and the corresponding information bit sequence, by independent (non
point very close to the (1.0, 1.0) MI point. According to our
does not intersect until a
point very close to the (1.0, 1.0, 1.0) MI point, which indicates that the
Fig. 5. EXIT curves of ACC−1 with doping ratio \( P_{ori} = 2 \) and \( C_0^{-1} \) with
rate 1/2. BSC crossover probabilities \( p_1 = 0.05 \) and \( p_2 = 0.06 \).

chain. (3) simulation. Under large enough SNR and sufficient
times of iterations, the trajectory can finally reach a point very
close to the (1.0,1.0,1.0) MI point, which indicates that the
original message, relayed by the forwarding nodes, can be
reconstructed perfectly.

Fig. 5 shows the EXIT curves and trajectory of the decoders
\( ACC^{-1} \) and \( C_0^{-1} \) where \( p_1 = 0.05 \), \( p_2 = 0.06 \) and \( P_{ori} = 2 \). After several iterations, the trajectory achieve (1.0,1.0) point
and the originator’s information be recovered completely. It
should be emphasized here that Fig. 5 indicates the case where
two HI’s perform as many iterations as no more gain in mutual
information can be achieved. However, even without full
iterations of two HI loops, which results in even smaller value
of mutual information after the \( a \ posteriori \) LLRs combining,
the EXIT curves of \( ACC^{-1} \) and \( C_0^{-1} \) do not intersect until a
point very close to the (1.0, 1.0) MI point. According to our
simulations, mutual information of 0.73 after the \( a \ posteriori \)
LLRs combining which is the case of non-full iterations over
two HI’s, can still keep the tunnel open.

V. SIMULATION RESULTS

BER performances of the proposed system with a two
representative value pairs of \( p_1 \) and \( p_2 \) are shown in Fig. 6 and
Fig. 7. In our simulations, we set the frame length at
100000 bits. 30 HI’s and 5 VI’s are performed in the joint
decoding part at the FD. After LLRs combing, we performed
6 and 3 HI’s for the \( p_1 \) and \( p_2 \) value sets, shown in Fig. 6 and
Fig. 7, respectively.

It is found from the BER simulation results that, we can
achieve clear turbo cliff over the AWGN channel. Also, we
can see that, a performance gain of about \( 2 - 3 \) dB can be
achieved by performing VI. The larger the gain, the smaller
the \( \rho \) value, in which case the two forwarding nodes are highly
correlated.2

VI. RATE OPTIMIZATION

Since the gaps between the EXIT curves of \( ACC^{-1} \) and
\( C_0^{-1} \) are very large as shown in Fig. 5, we can increase the rate
of \( C_0 \) by, for example using punctured convolutional codes to
achieve better matching of the two EXIT curves. As shown in
Fig. 8, it is found that even we can increase the coding rate of
\( C_0 \) from 1/2 to 2/3, still it can possible to achieve arbitrarily
low BER. However, since the gaps between two EXIT curves
is still not very small, reducing the gap, which is related to
the optimal code design issue, is left as future study.

In fact, we do not provide in-depth information theoretic
considerations on the relationship between the code rates and
Hamming distortion [12] in this paper. However, when we
seek for the optimal rate allocations to the codes used by
the originator and forwarding nodes, we have to first identify

\[ \text{BER} = \begin{cases} 0 & \text{if } C_0^{-1} \text{ and } ACC^{-1} \text{ intersect} \\ 1 & \text{otherwise} \end{cases} \]

\[ \text{Rate} = \begin{cases} 1/2 & \text{if } C_0^{-1} \text{ and } ACC^{-1} \text{ intersect} \\ 2/3 & \text{otherwise} \end{cases} \]

2The CEO problem belongs to distributed lossy compression problem in
Network Information Theory, and the limits for this problem, are not yet
known, except for some special cases.
the relationship in the information theoretic framework of the CEO problem. Especially, the rate of $C_0$ should be determined by the rate-distortion function, which takes into account the Hamming distortion and allocated rate to the nodes in the network, in general. It is expected that the more forwarding nodes involved, still error-free communication is possible with a high rate of $C_0$, and ultimately, we may be able to eliminate $C_0$.

VII. CONCLUSION

In this paper, we examined coding and decoding strategies on the issue of WMNs from the CEO problem viewpoint. In WMNs, the energy and spectrum efficiencies should be optimized as the whole network rather than an assembly of many P2P connections. Each forwarding node is small transceiver with low power consumption, where energy is very scarce. We thereby proposed a very simple strategy at the forwarding nodes and a joint decoding scheme by exploiting the correlation knowledge among intra-links at the FD. Even though errors are detected in the signals received by the forwarding nodes, they are correlated because of coming from the same originating node. Therefore, utilization of the Slepian-Wolf theorem allowing distorted source recovery should be the theoretical basis of the WMNs transmission chain design. The simulation results show that we can achieve roughly $2-3$ dB gain compared with separately decoding scheme. By optimizing the code parameters, close-limit BER performance can be expected, which belongs to the issue of the standard EXIT matching problem. More fundamental question is that how the relationship between the code rates and Hamming distortion can be formulated, and how it can be solved with the aim of their applications to WMNs. This is left as future study.

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