

A Novel Cooperation-Based Network Coding Scheme for Walking Scenarios in WBANs

Hongyun Zhang^{*†}, Farzad Safaei^{*}, Le Chung Tran^{*}

^{*}School of Electrical, Computer and Telecommunications Engineering

Faculty of Engineering and Information Science, University of Wollongong, NSW 2522, Australia

Email: hz697@uowmail.edu.au, (farzad,lctran)@uow.edu.au

[†]College of Computer, National University of Defense Technology

410073, Changsha, Hunan, China

Abstract

In Wireless Body Area Networks (WBANs), the tradeoff between network throughput and energy efficiency remains a key challenge. Most current transmission schemes try to cope with the challenge from the perspective of general Wireless Sensor Networks (WSNs), which may not take the peculiarities of WBAN channels into account. In this paper, we take advantage of the correlation of on-body channels in walking scenarios to achieve a better trade-off between throughput and energy consumption. We first analyze the characteristics of on-body channels based on realistic channel gain datasets, which are collected by our customized wireless transceivers in walking scenarios. The analytical results confirm the rationale of our newly proposed transmission scheme A3NC, which explores the combination of the Aggregative Allocation (AA) mechanism in MAC layer and the Analog Network Coding (ANC) technique in PHY layer. Both theoretical analyses and simulation results show that the A3NC scheme achieves significant improvement in upload throughput and energy efficiency, compared to the conventional approaches.

Index Terms

Network coding, Energy efficiency, Cross-layer, Scheduling, Walking scenario, Wireless Body Area Networks.

I. INTRODUCTION

IN recent years, with the technological advancements in wireless communication, Micro-Electro-Mechanical Systems (MEMS) and integrated circuits, WBANs have become a more practical and promising technology [1]. However, WBANs have strict energy constraints as frequent change of batteries is inconvenient or even infeasible. Besides, sensory devices are usually placed either in or on the human body, thus the maximum radiated transmission power is restricted to comply with the limitation of Specific Absorption Rate (SAR) of local regulatory bodies (the limitation of Federal Communications Commission is 1.6 W/Kg). Meanwhile, up to 10 Mbps throughput should be offered to satisfy the required set of entertainment and healthcare services [2]. It is a challenging task to maintain a high throughput while fulfilling the specific energy efficiency requirement of WBANs.

Propagation paths in WBANs can experience fading due to different reasons, such as energy absorption, reflection, diffraction, shadowing by the body, and body postures [3]. All these unique features lead to high packet losses. Relay nodes, if deployed outside of the human body, might possess better channels and less stringent energy limitations. Therefore, cooperative communication [4] has received considerable interest in recent years [5]–[8]. Initial solutions apply the existing methods, which were proposed for general Wireless Sensor Networks (WSNs), to WBANs. These schemes may not be optimized for the unique challenges and specific characteristics of WBANs, such as spatial correlation of channels, stringent trade-off between energy requirement and throughput gains, and Quality of Service (QoS) requirements.

Due to the broadcast nature of wireless networks, Network Coding (NC) [9], which enables nodes to code or mix packets (or symbols) before forwarding, is a potential method to produce a significant improvement in the throughput, reliability, manageability, and QoS of wireless networks [10], [11]. The applications of network coding in WBANs are still an emerging area and most existing works consider the expansions of conventional wireless network coding schemes. The works of [12]–[15] expanded the cluster-based scheme proposed in [16], which divides intermediate relay nodes into multiple clusters, and NC packets are transmitted between clusters. However, because only one-hop or two-hop star topologies are considered suitable for most WBAN systems [2], the feasibility of cluster-based schemes are limited. In [17], the authors proposed a practical NC implementations in the context of WBANs. The linear NC is performed in the source node to improve the energy efficiency. In [18], the relay first demodulates the received symbols from sensor nodes and then XORs them to generate a network coded symbol. In a similar way, the works in [19]–[22] also utilize the NC technology in the relay node to improve the transmission efficiency. The authors of [23] focus on the energy efficiency of cooperation transmission, decode-and-forward network coding (DF-NC) is proposed to aid relay cooperative transmission. The DF-NC scheme achieves a near optimal outage probability performance while maximizing the energy efficiency of WBANs by fixing the average number of transmissions per node. In [24], the authors explored the application of NC for the cloud-assisted ambient assisted living (AAL) environments. Random linear network coding (RLNC) technology is employed to the source node to improve the efficiency of retransmission. Some other works provided new perspectives to the harnessing of NC in WBANs. In [25], the cooperative compressed sensing approach is combined with RLNC technology to increase the energy efficiency of WBANs. However, most of the previous works did not consider the particular characteristics of WBAN channel when employing the NC technology.

One particular feature of WBANs is that the signal attenuation is significantly affected by the shadowing of body tissues in addition to the distance between two devices. In other words, the movement and posture of the human body have a dramatic effect on the strength of the received signal. Motivated by the spatial cross-correlation of channels in walking scenarios [26], [27], we proposed a novel NC transmission scheme, named A3NC, in our previous work [28]. However, the simulation model in [28] was built based on a distance-based channel model, rather than adopting the realistic channel gain. Also, the theoretical performance analysis was not provided in [28]. Moreover, the cross-correlation of on-body channels was presumed to be significant in [28]. However, in [29], we performed a full-scale experiment-based analysis of on-body channels in the walking scenarios, and found that the on-body channels do not manifest meaningful cross-correlation, either negative or positive. The experimental measurements presented later in this paper show that the path loss discrepancy between the on-body channels, rather than the cross-correlation, is significant, and it provides a more suitable basis for the new proposed scheme A3NC.

The contributions of this paper are as follows:

- The real channel gain datasets in walking scenarios are collected by our customized portable wireless transceiver. In-depth analyses of these experimental datasets confirm the basis of our newly proposed A3NC transmission scheme.
- Instead of the distance-based channel model used in [28], the realistic channel gain datasets collected from our measurement campaign are imported into the IEEE 802.15.6-compatible simulation model to evaluate the performance of A3NC. Simulations show that the proposed A3NC achieves a better performance from upload throughput, energy efficiency and throughput balance perspectives.
- To deepen the understanding of all key parameters that affect system performance, we provide the mathematical analyses from all three perspectives, namely upload throughput, energy efficiency and throughput balance.

The rest of the paper is organized as follows. In Section II, the system model is presented. Then, the characteristics of on-body channels in the walking scenarios are analyzed in Section III. Section IV details the proposed A3NC scheme. In section V, the mathematical analyses are provided. Performance evaluation results based on the realistic channel gain datasets are presented in Section VI. Finally, Section VII concludes the paper.

II. SYSTEM MODEL

In this section, we first present the network model for typical walking scenarios. Then channel models, including the on-body channel and relay-related channel, are introduced. Finally, the notations utilized throughout this article are presented.

A. Network Model

We consider the walking scenarios where two sensors (SN_1 and SN_2) mounted on the wrists communicate with a hub, and a relay node deployed outside the human body is optional. As the hub is expected to be larger than a sensor node, the torso is considered to be a preferable part to mount the hub. In this paper, we consider two typical device deployments as depicted in Fig. 1, where the hub is placed on the abdomen (attached to the belt buckle) or on the back collar. The main task for the sensors is to continually upload monitoring data to the hub with or without the help of the relay node. All wireless devices operate in the half-duplex mode, and Time Division Multiple Access (TDMA) is used to schedule the channel resource.

B. Channel Model

Some current works, including our paper [28], simply utilize the distance-based formula to quantify the strength of received signal power. However, the path loss of on-body channels in WBANs is affected by many factors, such as the absorption effect of human tissues and the mobility of the human body. In this work, the wireless channels in the walking scenarios are classified into two categories: on-body channel and relay channel (as shown in Fig. 2). Since only the sensors and the hub are deployed on the body, the channel between the sensors and the hub is named as on-body channel. On the other hand, the channel to or from the relay node is called relay channel.

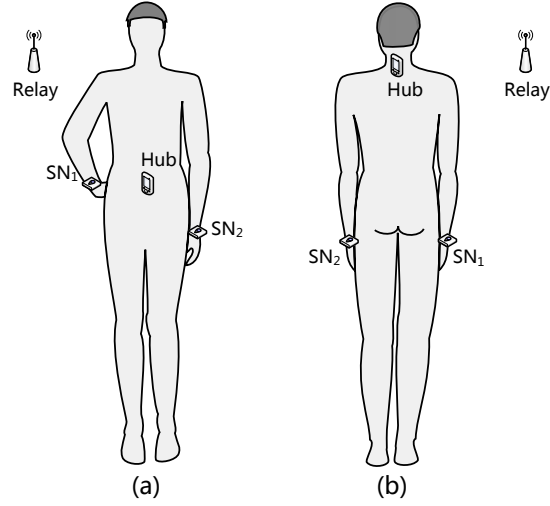


Fig. 1. Deployment of the sensors, relay and hub.

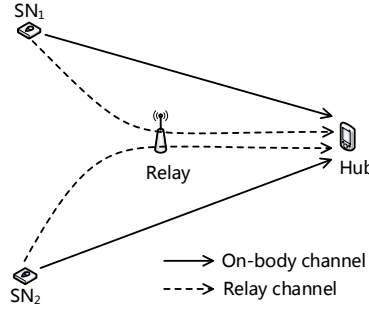


Fig. 2. The on-body channel and relay channel.

1) *On-body Channel Model:* Due to the high variability of on-body channels, neither distance-based nor other formula-based methods seem to be sufficient to describe the on-body channel condition, especially in the activity scenarios. Therefore, adopting channel gain datasets collected from the real walking scenarios to model the on-body channel is a better choice. To collect the realistic on-body channel gain data, we constructed wireless transceivers, as illustrated in Fig. 3. Each wireless transceiver consists of one radio module, one microcontroller and one MicroSD card. The main function of the transceivers is to transmit/receive continuous data packets to/from each other, thus facilitating the analysis of channel gains. Every testbed is composed of three printed circuit boards, and they stack on each other to make the system self-contained. More details about the hardware can be found in [29].

As demonstrated in [30] and [31], in narrowband communication environments, the on-body channels show prominent reciprocity, which means the channel profiles of downlink and uplink are around the same. Therefore, the channel gain can be estimated by the measurement of downlinks (from the hub to the sensors). The transmitter attached on the abdomen or the back collar continuously broadcasts sample packets to the receivers located on the two wrists, and the receivers record the channel gain data by time sequence. These datasets created from the experiment are utilized to model the on-body channels. More detail about these experiments and the analysis of channel gain datasets will be presented in the next section.

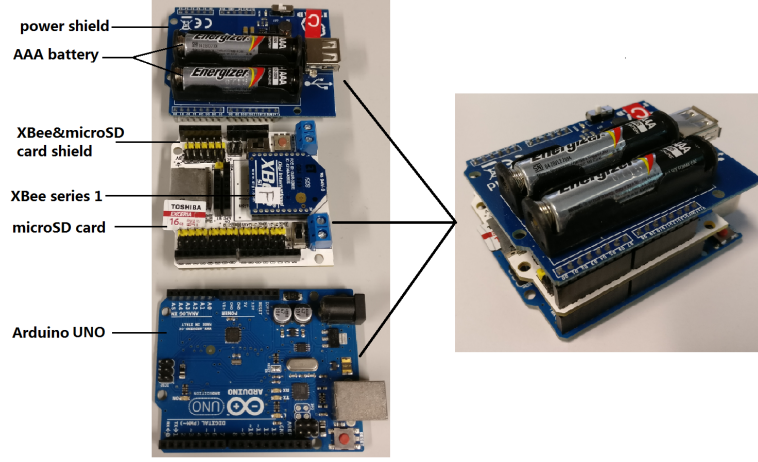


Fig. 3. Components of the wireless transceiver.

2) *Relay Channel Model*: In the context of WBANs, due to severe fading caused by body shadowing, the relay node may improve the transmission reliability effectively. As the off-body channel is considered to be much stable than the on-body channel, we assume the relay node is deployed outside the human body. In this work, the channel $SN_i - relay$ and channel $relay - hub$ are collectively called as “relay channel”. Besides, since the swing motions of two arm are symmetrical when the human is walking, the average packet delivery ratios (PDRs) tend to be the same for the two relay channels, i.e., $SN_1 - relay - hub$ and $SN_2 - relay - hub$. Accordingly, without loss of generality, we assume the average PDRs of the two relay channels are the same.

C. Notations

R refers to the transmission rate of upload data from the sensor to the hub, and the unit for it is Kbps, and the transmission rates of the two sensors are assumed to be the same. L_p denotes the length of one packet in Kbits. This paper considers beacon mode with superframe in the MAC layer, T_s denotes the length of one superframe, and T_c is the total time assigned to the two sensors and the relay node (if exist). Both T_s and T_c are in unit of second. β_i represents the average PDR of channel $SN_i - hub$ and β_r is average PDR of channel $SN - relay - hub$. As mentioned before, the two relay channels are assumed to have the same β_r . In the proposed A3NC schemes, two sensors share a common time period to upload monitoring data. Therefore, some packets may be lost due to the signal collision, and we denote the average packet error ratio (PER) caused by collision as e . Note that β_i is measured when the channel is only occupied by SN_i , without taking the packet losses caused by signal collisions into account. U_i denotes the actual upload throughput from SN_i to the hub with the influence of fading and collision. $\mathcal{P}_w$ is the working power of sensor node, and we assume the two sensors have the same working power.

III. CHARACTERISTICS OF ON-BODY CHANNELS

In this section, we first present the measurement setups for typical walking scenarios. Then the characteristics of the on-body channels are analyzed based on the datasets collected from these measurements.

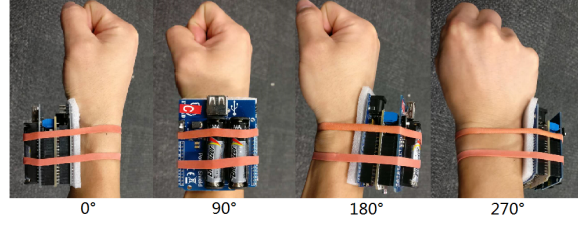


Fig. 4. Four directions of the transceiver on the wrist

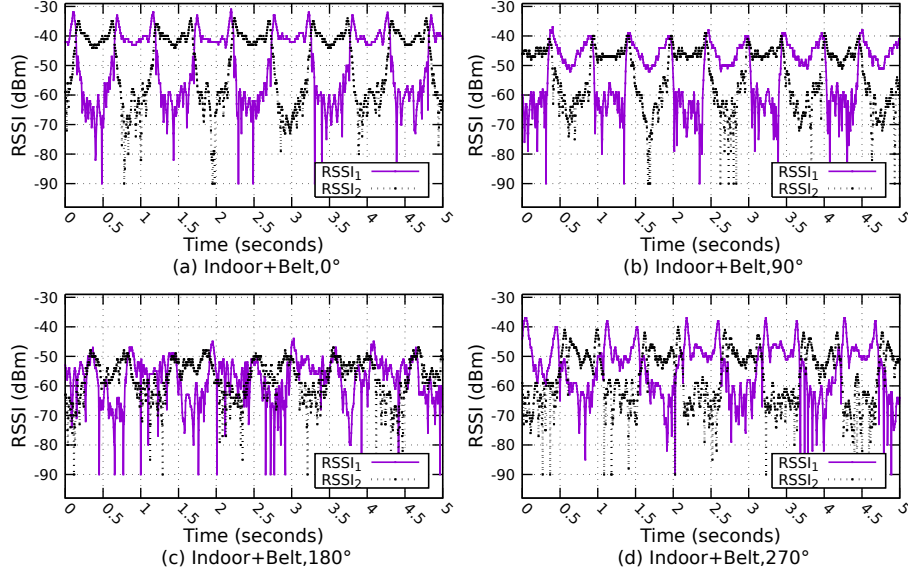


Fig. 5. RSSI vs. time in the “Indoor+Belt” scenario.

A. Measurement Setups

The measurements were conducted in both indoor and outdoor environments. The outdoor environment is an open oval field of about 13,000 square meters. The indoor environment is a hallway inside a building. Accordingly, there are four different walking scenarios generated based on the environment and the location of the hub. For the rest of the paper, four scenarios are named as “Indoor+Belt”, “Indoor+Collar”, “Outdoor+Belt” and “Outdoor+Collar”. Specifically, the transmitter located on the belt buckle or the back collar continuously broadcasts packets to the two receivers bound on the wrists with the transmit (Tx) power of 0 dBm and the sampling frequency of 200 Hz. Then the receivers record the RSSI (Received Signal Strength Indicator) value of received packets, the timestamp and the packet sequence number in the MicroSD card. Note that, as the Tx power is 0 dBm, the inverse of the RSSI is the path loss.

To explore the effect of the antenna direction and the shadowing effect of wrists, the portable transceivers are rolled around the wrist in four directions: 0°, 90°, 180° and 270°, as depicted in Fig. 4. Consequently, there are four different measurement setups (or sub-scenarios) in each scenario.

B. Channel Data Statistics

We first analyze the RSSI data for four different scenarios. Figs. 5-8 present typical time-varying RSSI data for the two links from the hub to the two sensors. $RSSI_1$ and $RSSI_2$ refer to the RSSI of the packets received by SN_1 and SN_2 , respectively.

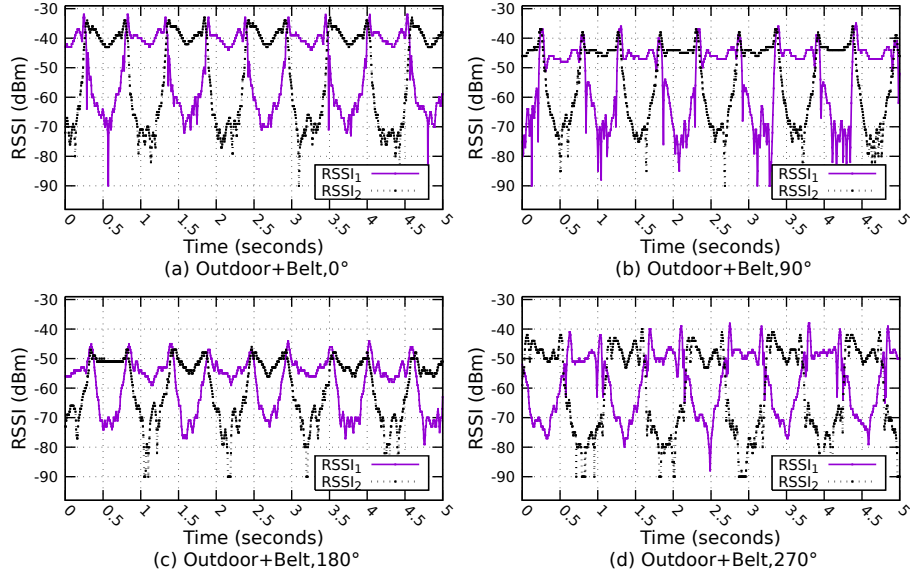


Fig. 6. RSSI vs. time in the “Outdoor+Belt” scenario.

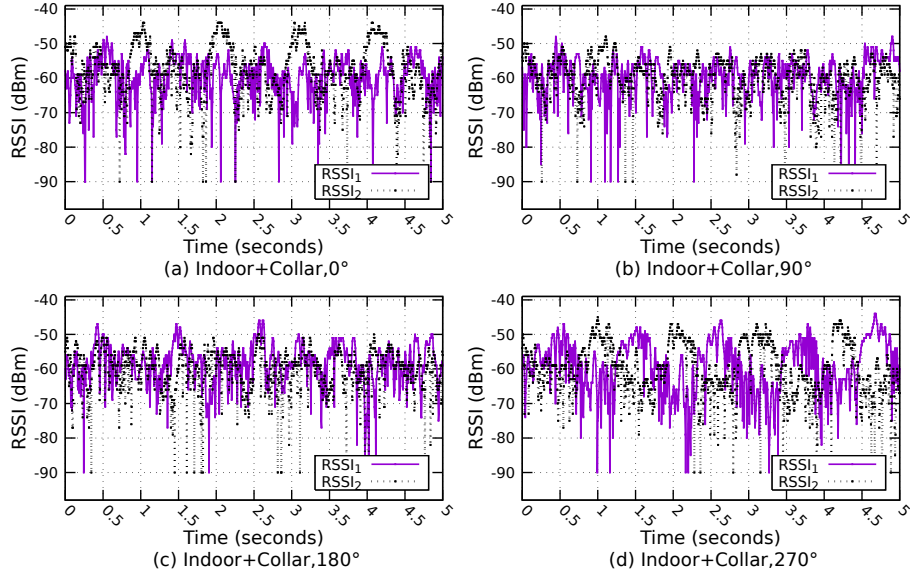


Fig. 7. RSSI vs. time in the “Indoor+Collar” scenario.

Observations that can be derived from the four figures are listed below.

- When the hub is on the “Collar”, the channel path losses are much bigger than the “Belt” cases. The main reason is that, in the “Collar” cases, the signal is affected not only by the shadowing of the torso but also the shadowing of the upper limbs
- The comparison between the indoor and outdoor environments shows that there exist more small-scale fluctuations in the indoor environment, resulting from the reflection from surrounding objects, such as desks, wall, and roof.
- The direction of transceivers also affects the signal attenuation. When the hub is on the abdomen, the 180° cases experience the biggest path loss. The reason is that when the transceivers are bound on the top of the wrists, the signal is not only affected by the shadowing of the torso but also the shadowing of the wrists themselves. Similarly, in the “Collar” cases, both 90° and 180° cases bring more shadowing from the upper limbs, leading to lower RSSI values.

Next, we analyze the cross-correlation between the two links from the sensors to the hub. The Pearson product-moment

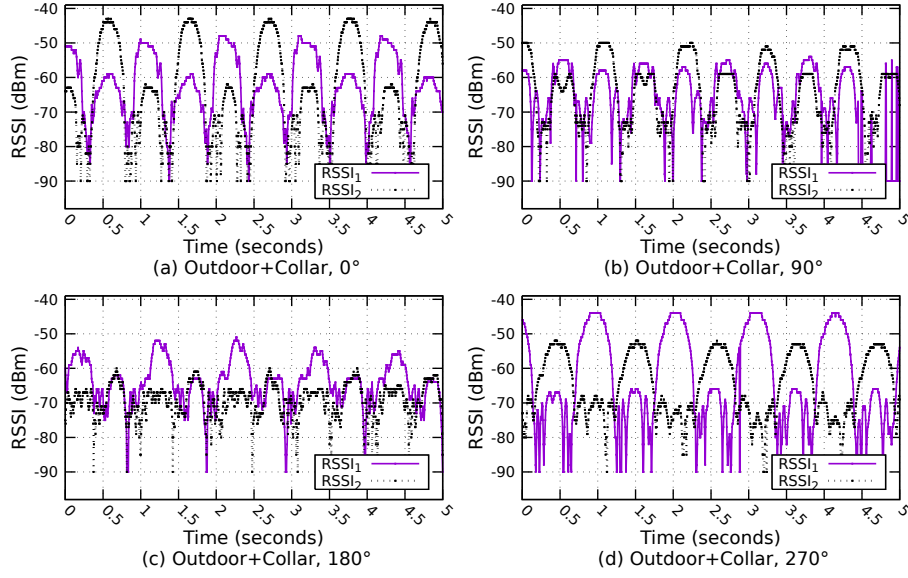


Fig. 8. RSSI vs. time in the “Outdoor+Collar” scenario.

TABLE I
CROSS-CORRELATION BETWEEN TWO LINKS.

	Indoor		Outdoor	
	Belt	Collar	Belt	Collar
0°	-0.45	-0.08	-0.42	-0.27
90°	0.15	0.00	0.43	0.14
180°	0.20	0.06	0.22	0.09
270°	-0.27	-0.26	-0.30	-0.44

correlation coefficient (PCC) [32] is adopted to evaluate the cross-correlation.

$$\rho_c = \frac{\sum_{n=1}^N (x(n) - \bar{x})(y(n) - \bar{y})}{\sqrt{\sum_{n=1}^N (x(n) - \bar{x})^2} \sqrt{\sum_{n=1}^N (y(n) - \bar{y})^2}} \quad (1)$$

where ρ_c is the correlation coefficient between $x(n)$ and $y(n)$, $\bar{x}$ and $\bar{y}$ are the mean path loss values of the link $SN_1 - hub$ and $SN_2 - hub$, respectively. $N=12,000$ is the number of samples in each measurement.

Table I presents the correlation coefficient for different experimental setups. It is found that the walking scenarios exhibit relatively small spatial cross-correlation coefficients, as the spatial cross-correlation is generally considered to be significant when its absolute value is 0.7 or greater. Further, the cross-correlations vary dramatically with the network deployment, including the placement of the hub and sensors, surrounding environment and the direction of transceivers. When the radio device rolls around the wrist, i.e., 0°, 90°, 180° and 270°, the PCC also varies accordingly. The differences mainly result from the change of antenna direction and the change of shadowing from the arms. Especially for the “Collar” cases, the significant shadowing coming from both torso and upper limbs outweighs other effects, leading to a relatively insignificant PCC.

The cross-correlation between different links is considered to be an instructive parameter for the design of communication systems. However, a small absolute value of PCC does not necessarily mean that two links are entirely independent. Take

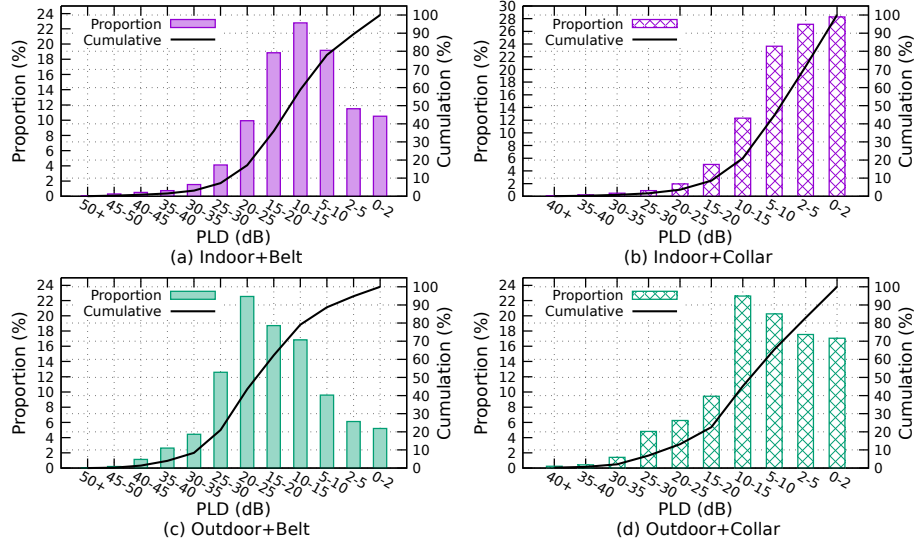


Fig. 9. The distribution of the PLDs.

the sub-scenario of Fig. 6(c) as an example. The PCC of the two channels is 0.22, but the two channels are obviously not independent. In this paper, we focus on a simple but important parameter, namely the path loss discrepancy (in dB), referred to as PLD, to reflect the actual difference between two channel gains.

As shown in Fig. 9, the probability distributions of the PLD are different for the four scenarios, but they all show a high proportion for the case when the PLD is greater than 5dB. Besides, the PLD medians for four experimental setups from Fig.9(a) to Fig.9(d) are 15dB, 7dB, 21dB and 11dB respectively. In the outdoor environment, the PLD tends to be greater than that in the indoor environment. The main reason is that the reflections from surrounding objects in the indoor environment narrow the gap between the RSSI of the two channels. Besides, the “belt” cases experience a bigger PLD than the “collar” cases, mainly because the shadowing effect is stronger in the latter.

Assuming that the environmental noise is relatively small, the PLD is the major constituent of the overall SNR (Signal-to-Noise Ratio). Meanwhile, SNR is a crucial parameter for the bit error ratio (BER). The relationship between SNR and BER has been investigated intensively in the literature [33], [34]. When $\text{SNR} > 7\text{dB}$, the system achieves a BER of lower than 1×10^{-2} for the most low-order modulation schemes, which may be considered to be sufficient for practical implementation. Accordingly, if two sensors located on the wrists concurrently broadcast packets to the hub on the torso in walking scenarios, the hub can restore at least one signal with a high probability. In other words, although the two sensors transmit simultaneously, the probability of packet collisions is small. Consequently, the observed PLD in walking scenarios may be sufficient to justify the rationale for simultaneous transmission from the two sensors to the hub, which is the core idea of our proposed A3NC scheme. The A3NC scheme will be detailed in the next section. More accurate quantitative mathematical analyses of this proposed scheme will be provided in Section V.

IV. JOINT ANALOG NETWORK CODING AND CHANNEL ALLOCATION

In A3NC, the aggregative allocation (AA) mechanism and the ANC [35] technique are incorporated to improve the system throughput and energy efficiency in walking scenarios. AA is the implementation method of simultaneous transmission in

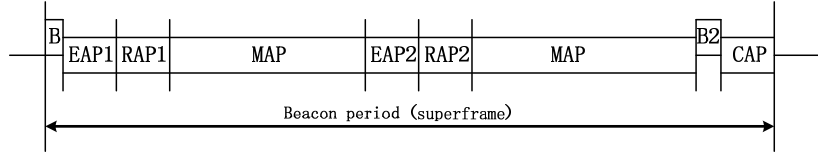


Fig. 10. Superframe structure of IEEE 802.15.6.

IEEE 802.15.6. ANC technique mainly focuses on the problem of energy efficiency, which is a crucial performance metric for WBANs. These two components of the proposed A3NC scheme are described below.

A. Aggregative Allocation

As mentioned in the previous section, the significant PLD in the walking scenarios is the motivation for simultaneous transmissions of different source nodes. The proposed AA mechanism aims to make the simultaneous transmission compatible with the IEEE 802.15.6 standard. Therefore, it is instructive to introduce the structure of the IEEE 802.15.6 superframe briefly. As depicted in Fig. 10, a superframe is divided into Exclusive Access Phases (i.e., EAP1 and EAP2), Random Access Phases (i.e., RAP1 and RAP2), Managed Access Phases (MAP) and a Contention Access Phase (CAP) [2]. In EAP, RAP and CAP periods, nodes contend for the channel allocation using either CSMA/CA (Carrier sense multiple access with collision avoidance) or slotted Aloha access procedure. In MAP periods, scheduled, unscheduled and improvised access methods are possible. The scheduled access procedure is generally used for applications that involve periodic monitoring.

In this paper, only the RAP1 and one MAP are adopted to compose the superframe in the beacon-enable mode. Compared to the contention-based methods, TDMA channel access method provides better energy efficiency. Therefore, TDMA method is utilized to upload the data from the sensors and the relay to the hub. As suggested by the IEEE 802.15.6 standard, the TDMA time slots should be assigned to the MAP duration. This is the reason that the MAP is utilized to upload the data from the sensors or the relay to the hub. As for RAP1 period, it precedes MAP to complete the transactions of the connection request and assignment on the basis of CSMA/CA method, which is the suggested approach for RAP in IEEE 802.15.6. These transactions occurred in the RAP1 period are essential for the AA mechanism, as they assign a shared transmission period to a pair of sensors with significant PLD and facilitate the cooperation of relay (The necessity of relay node will be discussed in Section IV.B). Besides, the one-periodic allocation is adopted, where the assigned intervals are reoccurring in every beacon period (superframe). The details of the assignment transactions are illustrated in Fig. 11.

The transactions depicted in Fig. 11 occur in the RAP1 phase. Initially, the hub broadcasts the beacon frame to the sensors (SN_1 and SN_2) and relay, and all receivers contend for the channel access by a CSMA/CA mechanism. Suppose that SN_1 randomly chooses the smallest backoff value. SN_1 transmits the connection request packet (Pkt_{CR}) to the hub to request an uplink interval in the MAP. Then, the hub sends a connection allocation packet (Pkt_{CA}) to SN_1 to allocate a scheduled uplink interval (SUI). The processes of allocating a SUI to SN_2 and the relay are similar. In addition, the relay's scheduled downlink interval (SDI) is allocated to the same period with the sensors' SUI in order to receive the packets from the sensors, which will be detailed in the next subsection. Note that handling the received Pkt_{CR} in the hub is the key process for the AA mechanism, which is also marked as three blue rectangles in Fig. 11.

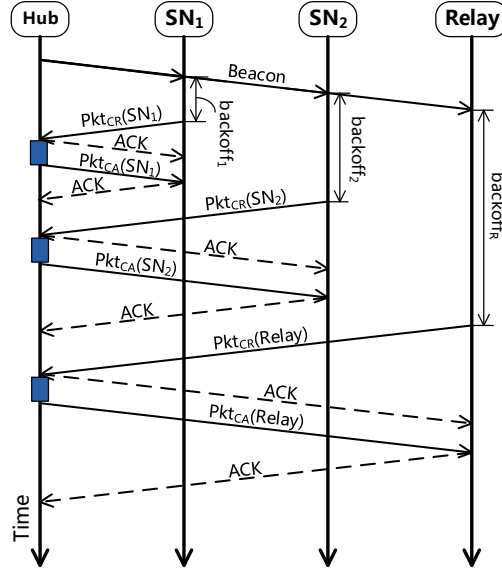


Fig. 11. Transaction procedure of A3NC in RAP1.

The detail of processing Pkt_{CR} in the hub is depicted in Algorithm 1. We add a new field item, named PID (Pair Identifier), in the header of Pkt_{CR} to assist the relay cooperation. More specifically, the two correlated sensors and their relay have the same PID value, and the SDI of the relay is assigned to overlap with the SUI of the two sensors to receive the packets from the two sensors. $NID_{assigned}$ denotes the NID (Node Identifier) assigned to the node. The hub maintains two independent maps: Map_{NID} and Map_{PID} . The Map_{NID} contains the mapping between MAC address and NID , and the Map_{PID} maintains the relation between PID and SUI or SDI . As illustrated in Algorithm 1 (Lines 2-5), a dedicated NID is assigned to each node (sensor or relay) with the help of Map_{NID} . Consequently, correlated sensors with the same PID share the same SUI by searching Map_{PID} (Lines 7-13). If Pkt_{CR} is sent by a relay, the relay node will be allocated with the SDI overlapped with the SUI of the two sensors with the same PID . Meanwhile, an SUI is allocated to the relay to forward data to the hub (as detailed in Lines 15-21). Note that SDI is not essential for sensors since their primary task is uploading the monitoring data to the hub.

After channel allocation, two sensors are allocated to share the same SUI , and the SDI of the relay is overlapped with the SUI of the two sensors. From the perspective of the hub, the two correlated sensors can be looked as one virtual sensor node. Meanwhile, the beacon period is fixed. Consequently, a longer SUI can be allocated to the sensors, allowing the AA technology to improve the system throughput in walking scenarios. The mathematical analyses will be presented in the next section.

B. Cooperative Communication with Analog Network Coding

In this subsection, we first explain the reasons for introducing the relay node. Firstly, by adopting pure AA without employing relay nodes, one of the sensors experiences more severe fading compared to the other one, which is a waste of energy. In other words, AA may achieve the throughput improvement at the cost of additional energy consumption, which is strictly limited in WBANs. Secondly, in the case that two correlated sensors experience different packets delivery ratios (PDR), AA may cause

Algorithm 1: Uplink and downlink allocation

Input: Receive a Pkt_{CR} from radio layer

```

1 Get  $PID$ , MAC address, required time slots from  $Pkt_{CR}$ ;
2 if Find a record in  $Map_{NID}$  by MAC Address then
3   | Set  $NID_{assigned}$  by  $Map_{NID}$ ;
4 else
5   |  $NID_{assigned}$  = current free connected  $NID$ ;
6 end
7 if  $Pkt_{CR}$  comes from a sensor node then
8   | if Find a record in  $Map_{PID}$  by  $PID$  then
9     | Set  $SUI.startSlot$  and  $SUI.endSlot$  by the record;
10  | else
11    | Set  $SUI.startSlot$  and  $SUI.endSlot$  from the end of  $Map_{PID}$ ;
12    | Update  $Map_{PID}$ ;
13  | end
14 else
15   | if Find a record in  $Map_{SID}$  by  $PID$  then
16     | Set  $SDI.startSlot$  and  $SDI.endSlot$  by the record;
17     | Set  $SUI.startSlot$  and  $SUI.endSlot$  from the end of  $Map_{PID}$ ;
18     | Update  $Map_{PID}$ ;
19   | else
20     | Discard  $Pkt_{CR}$ ;
21   | end
22 end
23 Construct a  $Pkt_{CA}$  based on  $NID_{assigned}$ ,  $SDI.startSlot$ ,  $SDI.endSlot$ ,  $SUI.startSlot$  and  $SUI.endSlot$ ;
24 Add  $Pkt_{CA}$  to sending buffer;

```

the sensor with the lower PDR to experience even a worse throughput performance than the conventional TDMA scheme. These two issues can be addressed effectively by deploying a relay node.

Since two sensors broadcast their signals concurrently, the traditional decode-and-forward (DF) relay strategy is impractical. Similarly, the digital network coding (DNC) technology, which requires the relay node to decode and re-encode the packets in the MAC layer, is also unrealistic. Given that concurrently transmitted packets from two sensors are mixed naturally at PHY layer and the hub can receive at least one sensor's packet with a high probability, physical-layer network coding (PLNC) technology may be a better choice. In our proposed A3NC, the analog network coding (ANC) technology originally proposed in [35] is adopted. The relay node simply amplifies-and-forwards the mixed signal to the hub without any decoding, and the hub undertakes the decoding using the mixed signal and one original signal. Perfect synchronization is not required in the ANC technology. Besides, since decoding of packets is not required, the complexity of the relay node is reduced and might be beneficial for the communication security.

Similar to [35], the MSK (Minimum Shift Keying) is adopted as an example modulation scheme to explain the main principle of ANC, for the reason that it has a simple demodulation algorithm as well as good bit-error properties. Fig. 12 illustrates how to incorporate ANC with cooperative communication. The cooperation procedure can be divided into two phases, which correspond to the SUI of the two sensors (also identical to the SDI of the relay) and the SUI of the relay. As depicted in

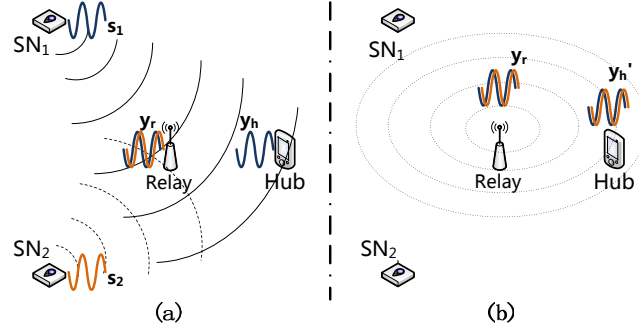


Fig. 12. Cooperative communication with ANC. (a) Phase 1: sensors' *SUI*. (b) Phase 2: relay's *SUI*.

Fig. 12 (a), the signals transmitted from SN_1 and SN_2 , respectively, can be represented as

$$s_1[n] = A_s[n]e^{i\theta_s[n]} \quad (2)$$

$$s_2[n] = B_s[n]e^{i\phi_s[n]} \quad (3)$$

where $A_s[n]$ and $B_s[n]$ are the amplitudes of the n -th sample, and $\theta_s[n]$, $\phi_s[n]$ are their phases.

If the signal from SN_2 (i.e., s_2) experiences a severe fading and the hub successfully receives the signal from SN_1 , the received signal at the hub, $y_h[n]$, can be represented as

$$y_h[n] = hA_s[n]e^{i(\theta_s[n]+\gamma)} + w(n) \quad (4)$$

where $he^{i\gamma}$ is the complex coefficient for channel $SN_1 - \text{hub}$ and $w(n)$ is the additive Gaussian noise. For brevity, *noise* term will be ignored in our subsequent formulas.

Meanwhile, the relay node receives and stores the mixed signal $y_r[n]$ from two sensors as follows:

$$y_r[n] = h_1A_s[n]e^{i(\theta_s[n]+\gamma_1)} + h_2B_s[n]e^{i(\phi_s[n]+\gamma_2)} \quad (5)$$

where h_1 and γ_1 are the channel gain and phase shift respectively for the channel $SN_1 - \text{Relay}$, and h_2 and γ_2 are those for the channel $SN_2 - \text{Relay}$.

In the second phase (i.e., the *SUI* of the relay node), which is shown in Fig. 12(b), the relay node amplifies the ANC mixed signal $y_r[n]$ and forwards it to the hub. Since the relay node only amplifies and forwards $y_r[n]$, for simplicity, we assume the received signal at the hub is the same as the mixed signal at the relay, which can be presented as

$$y'_h[n] = A[n]e^{i\theta[n]} + B[n]e^{i\phi[n]} \quad (6)$$

where $A = h_1A_s$, $B = h_2B_s$, $\theta[n] = \theta_s[n] + \gamma_1$, and $\phi[n] = \phi_s[n] + \gamma_2$. In the complex plane, $y'_h[n]$ is a sum of two vectors, which have lengths A and B respectively. As shown in Fig. 13, the two vectors lie on two circles with radii A and B . As

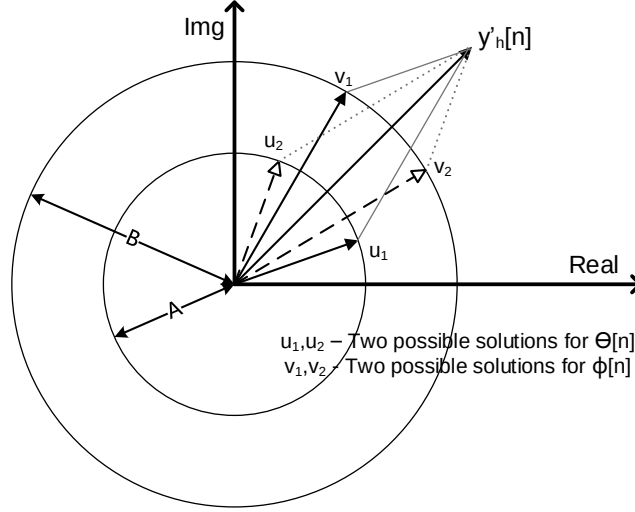


Fig. 13. Two possible solutions for the phase computation.

proved in [35], there exist two solutions for the pair $(\theta[n], \phi[n])$

$$\theta[n] = \arg(y'_h[n](A + BD \pm iB\sqrt{1-D^2})) \quad (7)$$

$$\phi[n] = \arg(y'_h[n](A + AD \mp iA\sqrt{1-D^2})) \quad (8)$$

where $D = \frac{|y'_h[n]|^2 - A^2 - B^2}{2AB}$, $|y'_h[n]|$ is the norm, and $\arg$ is the angle of the complex number. Note that for each solution $\theta[n]$, there is a unique solution for $\phi[n]$. Specifically, if $\theta[n] = \arg(y'_h[n](A + BD + iB\sqrt{1-D^2}))$, the corresponding solution of $\phi[n]$ is $\arg(y'_h[n](A + AD - iA\sqrt{1-D^2}))$. In addition, the amplitude of the two signals (i.e., A and B) can be estimated by the received signal [35].

The next step is to estimate the phase difference between the signals from two sensors, i.e., $\theta[n+1] - \theta[n]$ and $\phi[n+1] - \phi[n]$. Corresponding to two potential phase pairs at each sample time, there are four possible pairs of phase differences

$$(\Delta\theta_{xy}[n], \Delta\phi_{xy}[n]) = (\theta_x[n+1] - \theta_y[n], \phi_x[n+1] - \phi_y[n]) \quad \forall x, y \in \{1, 2\} \quad (9)$$

Since the channel of $SN_1 - \text{hub}$ is in a “good” condition, the hub knows the signal transmitted from SN_1 , $y_h[n]$ (cf. Eq. (4)). Thus, the hub knows the phase difference $\Delta\theta[n]$. Next, the hub picks the $\Delta\theta_{xy}[n]$ that produces the smallest deviation to the $\Delta\theta[n]$ as the optimal one, and the matching $\Delta\phi_{xy}[n]$ is the optimal phase difference for the signal from SN_2 . In the last step, the hub obtains SN_2 's bits based on these estimated phase differences.

Since the focus of this paper is the spatial correlation of on-body channels and the combination of AA and ANC, many other decoding details of ANC, such as the estimation of amplitudes A and B and the alignment of pilot sequences, are omitted in the above example. More details of the ANC technology can be found in [35].

V. PERFORMANCE ANALYSIS

In this section, the performance of the proposed scheme is analyzed and compared to the existing schemes from three perspectives: total throughput, energy efficiency and throughput balance. The AA mechanism and ANC cooperation are two

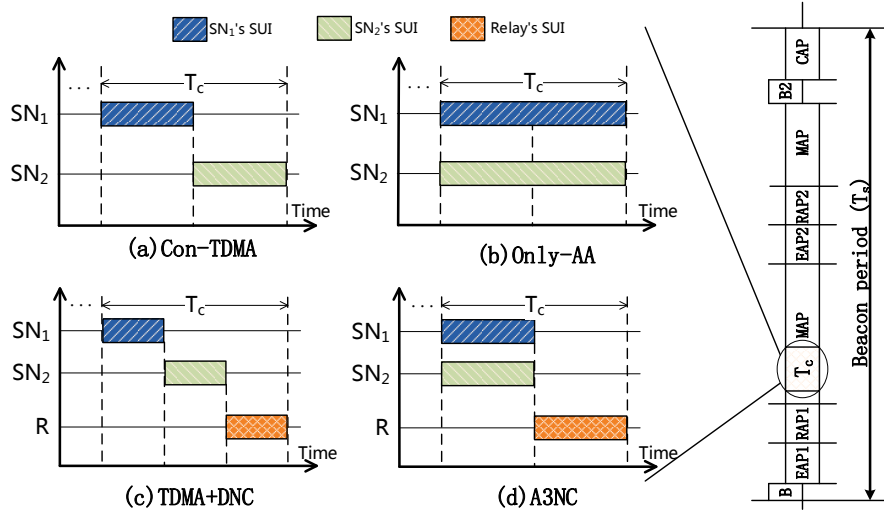


Fig. 14. Slot schedules of four schemes.

key components of the proposed A3NC scheme. Three alternative schemes that do not employ both AA and ANC are considered for comparison, and their time slot schedules are depicted in Fig.14.

As shown in Fig. 14, suppose a certain period T_c is allocated to the two sensors and the corresponding relay node. In the conventional TDMA scheme, denoted as Con-TDMA (Fig. 14(a)), T_c is split into two for the *SUI* of the two sensors. Whereas, when AA mechanism is employed without the cooperative communication part of A3NC, denoted as Only-AA (Fig. 14(b)) both sensors share the whole T_c concurrently. Next, there are two different approaches to assign the *SUI* of the relay. First, when the relay node forwards the digital network coding (DNC) packets to the hub in the third time interval, and each of the two sensors possesses an exclusive *SUI*. This approach is named as TDMA-DNC, in which the T_c is divided into three identical parts. The second one is our proposed scheme A3NC (Fig. 14(d)), the two sensors are assigned to a shared *SUI*, and the ANC is adopted as the relay technology to improve the performance.

A. Throughput Analysis

Before analyzing the upload throughput of the four schemes, it is necessary to examine e in more detail. As introduced in Section II, e denotes the packet error ratio (PER) caused by signal collision, which only occurs when two sensors perform simultaneous transmissions in the MAP duration, so e only exist in Only-AA and A3NC. As will be presented in the next section, e in the typical walking scenario would be small enough (about 0.022) to be ignored. However, in order to explore the performance of A3NC in other situations, rather than being limited to walking scenarios, we consider e as a variable parameter in both mathematical and simulation analyses. Besides, we assume e for the two sensors to be identical due to symmetry.

1) *Conventional TDMA*: We first consider the Con-TDMA scheme without the cooperation of the relay node. In this scheme, every sensor is scheduled with a dedicated uplink interval. The expectation of upload throughput of SN_1 , denoted as $E(U_1)$ (Kbps), can be calculated as

$$E(U_1)_{Con-TDMA} = (R \times \frac{1}{2}T_c \times \frac{1}{L_p} \times \beta_1 \times L_p) \times \frac{1}{T_s} = \frac{1}{2} \frac{RT_c\beta_1}{T_s} \quad (10)$$

Since only two sensors share the time interval, every sensor occupies $\frac{1}{2}T_c$ upload interval; $R \times \frac{1}{2}T_c \times \frac{1}{L_p}$ is the number of packets transmitted in one beacon period by SN_1 . ($R \times \frac{1}{2}T_c \times \frac{1}{L_p} \times \beta_1 \times L_p$) is the amount of bits received by the hub. The calculation for the upload throughput of SN_2 , i.e., $E(U_2)$, is similar. Thus, the expectation of the sum of upload throughput of the two sensors can be expressed as

$$E(U_1 + U_2)_{Con-TDMA} = \frac{1}{2} \frac{RT_c}{T_s} (\beta_1 + \beta_2) \quad (11)$$

2) *Only-AA*: In the Only-AA scheme, since two sensors transmit simultaneously, the packet errors caused by signal collisions should be considered. The actual PDRs for SN_1 and SN_2 are $(\beta_1 - e)$ and $(\beta_2 - e)$ respectively. Accordingly, the throughput of the two sensors can be expressed as

$$E(U_1 + U_2)_{Only-AA} = \frac{RT_c(\beta_1 - e)}{T_s} + \frac{RT_c(\beta_2 - e)}{T_s} = \frac{RT_c}{T_s} (\beta_1 + \beta_2 - 2e) \quad (12)$$

3) *TDMA+DNC*: The TDMA+DNC scheme is a classical approach that adopts DNC as the cooperative technology. As Fig. 14(c) shows, the relay node occupies the third time interval to transmit the network coding packets to the hub. Since the packet loss in the first two time slots may be recovered in the third time slot, the probability of effective cooperative communication should be taken into account to calculate the actual PDR (denoted as $\mathcal{B}_1$ and $\mathcal{B}_2$). If the result of whether the hub receives the packets from SN_1 is represented by a discrete random variable (D.R.V) X , then let $X = S$ denote the situation that the hub receives the packets from SN_1 successfully, and $X = F$ represents the failure of the hub to receive the packets from SN_1 . Similarly, D.R.V Y and R denote the situations for SN_2 and the relay node respectively. Hence, $\mathcal{B}_1$ and $\mathcal{B}_2$ can be calculated as

$$\mathcal{B}_1 = P_{X,Y}(S, F) + P_{X,Y}(S, S) + P_{X,Y}(F, S)P_R(S) \quad (13)$$

$$\mathcal{B}_2 = P_{X,Y}(F, S) + P_{X,Y}(S, S) + P_{X,Y}(S, F)P_R(S) \quad (14)$$

where $P_{X,Y}$ is the joint probability that considers both D.R.V X and Y , and P_R denotes the probability of whether the hub receives the packets from the relay node. It is worth noting that $\beta_1 = P_{X,Y}(S, F) + P_{X,Y}(S, S)$, $\beta_2 = P_{X,Y}(F, S) + P_{X,Y}(S, S)$ and $\beta_r = P_R(S)$. Accordingly, $\mathcal{B}_1$ and $\mathcal{B}_2$ can be expressed as

$$\mathcal{B}_1 = \beta_1 + (\beta_2 - P_{X,Y}(S, S))\beta_r \quad (15)$$

$$\mathcal{B}_2 = \beta_2 + (\beta_1 - P_{X,Y}(S, S))\beta_r \quad (16)$$

For a general wireless network, it is not straightforward to calculate the probability that the packets from two sensors are received by the hub, i.e., $P_{X,Y}(S, S)$. However, since we mainly consider the walking scenarios, $P_{X,Y}(S, S)$ can be approximated

as

$$P_{X,Y}(S, S) \approx \begin{cases} 0, & \beta_1 + \beta_2 \leq 1 \\ (\beta_1 + \beta_2) - 1, & \beta_1 + \beta_2 > 1 \end{cases} \quad (17)$$

We derive the above approximation based on the following observations. Compared to the walking cycle (about 1000 ms - see in Section III.B.), the superframe (about 100 ms) is relatively short. Moreover, the two upload intervals assigned to the two sensors are typically adjacent. So, based on the statistical results in Section III, it is reasonable to assume the PLD between the two channels remains significant and stable in one superframe. To make the approximation in Eq. (17) clear, suppose that each sensor only sends one packet in one superframe. Accordingly, in the case of $\beta_1 + \beta_2 \leq 1$, the hub tends to receive only one packet either from SN_1 or SN_2 with a high probability in one superframe, instead of receiving packets from both sensors simultaneously. On the other hand, in the case of $\beta_1 + \beta_2 > 1$, the received signal strength from the “bad” channel is still strong enough to decode, regardless of the big PLDs between the two channels. Note that the latter case is normally caused by the increase of Tx power of sensors. Therefore, based on Eqs. (15-17), the expected overall throughput can be obtained as

$$E(U_1 + U_2)_{TDMA+DNC} = \frac{1}{3} \frac{RT_c \mathcal{B}_1}{T_s} + \frac{1}{3} \frac{RT_c \mathcal{B}_2}{T_s} \approx \begin{cases} \frac{1}{3} \frac{RT_c}{T_s} (\beta_1 + \beta_2)(1 + \beta_r), & \beta_1 + \beta_2 \leq 1 \\ \frac{1}{3} \frac{RT_c}{T_s} ((\beta_1 + \beta_2)(1 - \beta_r) + 2\beta_r), & \beta_1 + \beta_2 > 1 \end{cases} \quad (18)$$

4) *A3NC*: As for the A3NC scheme, two sensors broadcast the data packets simultaneously in the first half period, and the ANC packet is relayed in the second half. This situation can be considered as an integration of Only-AA and TDMA+DNC schemes. When calculating the actual PDRs, denoted as $\tilde{\mathcal{B}}_1$ and $\tilde{\mathcal{B}}_2$, both the cooperative communication and the packet loss caused by signal collision should be taken into account. The actual PDRs for SN_1 and SN_2 are expressed as

$$\tilde{\mathcal{B}}_1 = (\beta_1 - e) + (\beta_2 - e)\beta_r \quad (19)$$

$$\tilde{\mathcal{B}}_2 = (\beta_2 - e) + (\beta_1 - e)\beta_r \quad (20)$$

The first parts of Eqs. (19-20) refer to the PDRs for the first half timeslot, and the second parts are the PDRs with the help of the relay node. Consequently, the upload throughput of the whole network can be written as

$$E(U_1 + U_2)_{A3NC} = \frac{1}{2} \frac{RT_c \tilde{\mathcal{B}}_1}{T_s} + \frac{1}{2} \frac{RT_c \tilde{\mathcal{B}}_2}{T_s} = \frac{1}{2} \frac{RT_c}{T_s} (\beta_1 + \beta_2 - 2e)(1 + \beta_r) \quad (21)$$

As a summary, we list some important observations by comparing the analytical results of the four schemes.

Observation 1:

If the packets error ratio caused by signal collision is below the quarter of $\beta_1 + \beta_2$, the throughput performance of Only-AA is better than that of Con-TDMA scheme. That is $E(U_1 + U_2)_{Only-AA} > E(U_1 + U_2)_{Con-TDMA}$, when $e < \frac{1}{4}(\beta_1 + \beta_2)$.

Observation 2:

The upload throughput performance of A3NC is always under that of Only-AA, because the PDR of the relay channel

TABLE II
THE UPPER BOUND OF THROUGHPUT.

	Con-TDMA	Only-AA	TDMA+DNC	A3NC
Expectation of throughput	$\frac{1}{2}M$	M	$\frac{2}{3}M$	M

cannot exceed one, i.e., $\beta_r \leq 1$.

$$E(U_1 + U_2)_{A3NC} \leq E(U_1 + U_2)_{Only-AA}$$

Observation 3:

The comparison between Con-TDMA and TDMA+DNC is complicated. If $\beta_1 + \beta_2 \leq 1$, we have

$$E(U_1 + U_2)_{TDMA+DNC} \begin{cases} > E(U_1 + U_2)_{Con-TDMA}, & \beta_r > \frac{1}{2} \\ \leq E(U_1 + U_2)_{Con-TDMA}, & \beta_r \leq \frac{1}{2} \end{cases}$$

If $1 \leq \beta_1 + \beta_2 \leq \frac{4}{3}$, and we denote $Q = \frac{\beta_1 + \beta_2}{4 - 2(\beta_1 + \beta_2)}$, the relationship between these two schemes can be expressed as

$$E(U_1 + U_2)_{TDMA+DNC} \begin{cases} > E(U_1 + U_2)_{Con-TDMA}, & \beta_r > Q \\ \leq E(U_1 + U_2)_{Con-TDMA}, & \beta_r \leq Q \end{cases}$$

If $\beta_1 + \beta_2 > \frac{4}{3}$, the throughput performance of TDMA+DNC cannot exceed that of Con-TDMA. That is

$$E(U_1 + U_2)_{TDMA+DNC} < E(U_1 + U_2)_{Con-TDMA}$$

Observation 4:

When $e < \frac{1}{6}(\beta_1 + \beta_2)$, A3NC achieves a better throughput performance than TDMA+DNC

$$E(U_1 + U_2)_{A3NC} > E(U_1 + U_2)_{TDMA+DNC}$$

It can be seen that both A3NC and Only-AA outperform the conventional schemes when the values of specific parameters (i.e. e , β_r and $\beta_1 + \beta_2$) satisfy certain conditions. Detail performance evaluation based on walking datasets will be provided in the next section.

In addition, if the packets losses due to signal collisions and failed cooperation are small, i.e., $e \approx 0$ and $\beta_r \approx 1$, we can derive the upper bounds for the four schemes in the walking scenarios. For brevity, denote $M = \frac{RT_c}{T_s}(\beta_1 + \beta_2)$. The upper bounds of upload throughput are depicted in Table II. We notice that the upper bounds for Only-AA and A3NC are the same, i.e., $\frac{RT_c}{T_s L_p}(\beta_1 + \beta_2)$, which achieves 2 and 1.5 throughput gain in comparison with Con-TDMA and TDMA+DNC respectively.

TABLE III
ENERGY EFFICIENCY COMPARISON.

	Con-TDMA	Only-AA	TDMA+DNC	A3NC
T_{active}	T_c	$2T_c$	$\frac{2}{3}T_c$	T_c
EE	$\frac{1}{2}N$	$\frac{1}{2}N$	N	N

B. Energy Efficiency Analysis

An effective MAC layer method for energy saving is prolonging sleep time of the sensors, which also means completing data transmission in a shorter active time. Generally, recharging and changing the battery in a hub are easier than in the sensors. The relay node is deployed out of human body. Hence, in this paper, the energy efficiencies of hub and relay are not considered. Moreover, due to the behaviors of hub, relay and sensor nodes are repeated in each beacon period, we take the active time in one beacon period as an example. Note that the active state of the radio includes both Tx (transmission) and Rx (reception) states. Assuming the working power of sensors remains the same for both Rx and Tx states, the energy efficiency (EE) in Kbits/Joule can be defined as the ratio of the amount of uploading data and the power consumed in one beacon period

$$EE = \frac{E(U_1 + U_2)T_s}{T_{active}\mathcal{P}_w} \quad (22)$$

where T_{active} is the active time of the two sensors in one beacon period, and $\mathcal{P}_w$ is the working power of the sensor. $E(U_1 + U_2)$ is the expected sum of the upload throughput of the two sensors, and it has been analyzed before. We now focus on the T_{active} , which can be calculated from Fig. 14, Similar to the analysis of throughput, if $e=0$, $\beta_r=1$ and denote $N = \frac{R(\beta_1 + \beta_2)}{\mathcal{P}_w}$, the upper bounds of EE are summarized in Table III. From Table III, the two cooperative schemes (TDMA+DNC and A3NC) achieve a twofold EE gain, compared to Con-TDMA and Only-AA.

Combining the results in Tables II and III, we find that the A3NC scheme not only achieves the highest throughput upper bound but also retains the best energy efficiency. In other words, the A3NC scheme provides a better trade-off between network throughput and energy efficiency.

C. Throughput Balance Analysis

The above analyses consider the total upload throughput of the two sensors. In this subsection, the throughput disparity between the two sensors (i.e., $E(U_1)$ and $E(U_2)$) is taken into account, instead of only considering their sum. The aim is to assess whether the proposed A3NC scheme may result in a severe throughput imbalance between the two sensors, for example, one sensor being completely starved. We define a new parameter called the balance factor (BF) to evaluate whether the transmission scheme can achieve a reasonable throughput balance between the two sensors. The balance factor BF is defined by the following equation

$$BF = \left| \log_2 \frac{E(U_1)}{E(U_2)} \right|$$

TABLE IV
BALANCE FACTOR COMPARISON.

	Con-TDMA	Only-AA	TDMA+DNC	A3NC
BF	$\log_2 \frac{\beta_1}{\beta_2}$	$\log_2 \frac{\beta_1 - e}{\beta_2 - e}$	0	0

where $E(U_1)$ and $E(U_2)$ are the expected throughput of SN_1 and SN_2 , respectively. If $BF = 0$, the two sensors achieve the same throughput, no matter whether β_1 is equal to β_2 or not. On the other hand, the larger the BF is, the bigger throughput disparity between the two sensors. Table IV summarizes the BF when the relay node achieves a perfect cooperation, i.e., $\beta_r = 1$.

As shown in Table IV, the BF s of the two cooperative schemes (TDMA+DNC and A3NC) are zero, but the BF s of other two non-cooperative schemes vary with the ratio of two actual PDRs. That is because, with the help of the relay, the hub can decode the packet sent through the “bad” channel by receiving a cooperative packet from the “good” channel, which balances the throughput of the two sensors. Therefore, TDMA+DNC and A3NC achieve a better throughput balance, compared to the schemes without relay nodes.

In summary, the proposed A3NC scheme achieves a better throughput in comparison with the schemes without AA method. Meanwhile, due to the contribution of relay node, the A3NC also achieves a better energy efficiency and throughput balance compared to the schemes without the relay.

VI. PERFORMANCE EVALUATION

In this section, simulation results will be provided to show the performance of the A3NC scheme in terms of total throughput, energy efficiency and throughput balance.

A. Simulation Model and Configurations

The protocol stacks from PHY layer to application layer are developed based on the Castalia framework [36] in OMNeT++. All the important default parameters of protocol stacks and hardware are listed in Table V. The parameters of the routing layer are not listed in the table because the routing layer in the model only forwards packets between the application layer and MAC layer. Besides, the initial energy of the sensor is set to 2430 Joules, which is the average energy of one CR2032 lithium button battery with 3.0 voltages and 225 mAh capacity. For the fairness of comparison, the SUI of the sensors and the SUI of the relay are set up to different values in different schemes to make sure that the overall time (T_c) remains 120 ms (as depicted in Fig. 14).

In this paper, both the path loss datasets and a simulated network model are utilized to construct channel-related parameters. First, the datasets collected from experiments are imported to simulate the variation of the path loss in time. However, these datasets only cover a limited number of walking scenarios, and the critical parameters, i.e., β_i , β_r and e , in each scenario are fixed. In order to verify the theoretical analyses results mentioned in Section V, a simulated network model should also be used to adjust β_i , β_r and e . The combination of the experimental datasets and simulated network model construct a more flexible simulation platform.

TABLE V
SIMULATION PARAMETERS.

Parameter	Value	Parameter	Value
Radio Layer			
dataRate	512 kbps	modulationType	DIFFBPSK
bandwidth	20 MHz	carrierFreq	2400.0 MHz
noiseFloor	-101 dBm	sensitivity	-91 dBm
CCAthreshold	-95 dBm	symbolsForRSSI	8 bits
Tx Power	-38 dBm	Tx state energy consumption	2.9 mW
Rx state energy consumption	3.1 mW	Idle state energy consumption	0.05 mW
MAC Layer			
beaconPeriodLength	160 ms	RAP1Length	40 ms
SlotLength	5 ms	pTIFS	0.03 ms
Data's ACKType	N_ACK	Control's ACK	I_ACK
others			
sensor's initialEnergy	2430 J	simulationTime	50 s

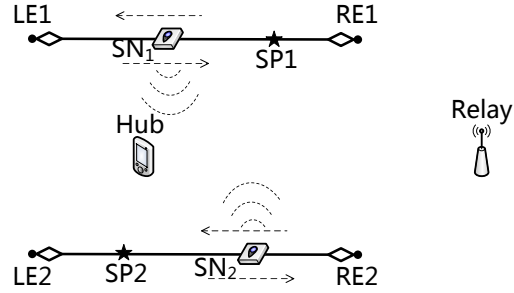


Fig. 15. Simulated network model in OMNeT++.

Fig. 15 presents the simulated network model. SP1, SP2 are the starting points for the two sensors. The reciprocating motion of the two sensors is simulated by importing the trace file, which records the path loss sequence during the walking activities. We now explain how to tune β_i , β_r and e . The “Outdoor+Belt” scenario with the transceivers bound in 0 degree is picked as the typical one, and the corresponding path loss dataset is imported into the network model. First of all, the average PDR for the channel $SN_i - hub$ (i.e., β_i) can be changed by modifying the Tx power. As shown in Fig. 16, when the Tx power of the two sensors decreases, the sum of β_1 and β_2 also diminishes. Moreover, due to the two sensors have the same Tx power, β_1 and β_2 are also approximately identical.

The average PDR of channel $relay - hub$ (i.e., β_r) can be tuned by setting up a parameter (named PDR_Relay) in OMNeT++ to modify the behavior of the relay node. At last, the average PER due to collisions (i.e., e) is controlled by changing the relative position between the two sensors’ start points (SP1 and SP2). From Fig. 15, intuitively, the interference between the two sensors is maximized (i.e., maximum e), if both SP1 and SP2 are in the left endpoint of their motion trajectories (i.e.

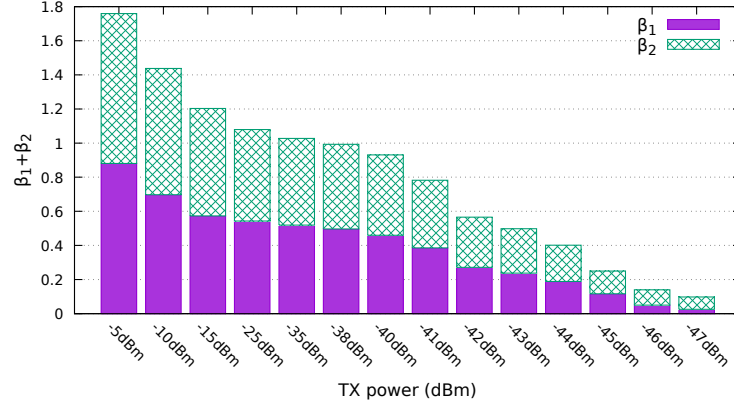


Fig. 16. $\beta_1 + \beta_2$ vs. TX power.

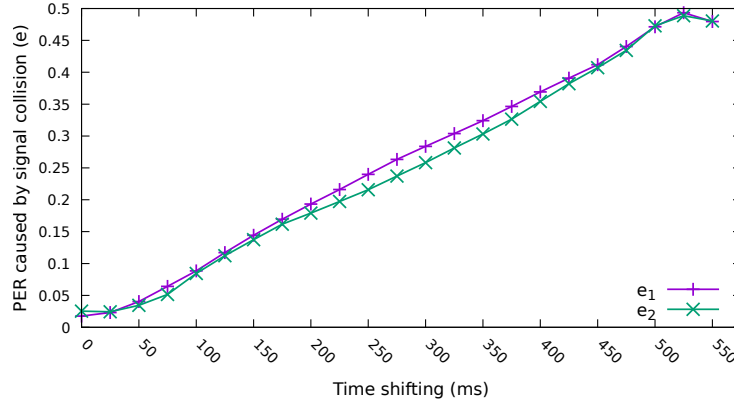


Fig. 17. e vs. Time shifting.

LE1 and LE2). But, if SP1 is on LE1 and SP2 is on RE2, which is a common situation in walking scenarios, e reaches its minimum value. Accordingly, we can change the value of e by changing the two start points of the sensors. Specifically, e can be changed by scrolling the path loss sequence of SN_2 while remaining the path loss sequence of SN_1 . As the sampling rate is fixed, the scrolling of path loss sequence can also be denoted as the time shifting compared to the normal arm swing in the walking scenarios.

Fig. 17 presents the curve of e versus time shifting, when the Tx power of the two sensors is set to -38 dBm and $\beta_1 + \beta_2$ is approximately one. The case of “time shifting = 0” corresponds to the normal swing motion of arm in walking scenarios. As shown in Fig. 17, the average PER due to collisions for normal walking scenarios is about 0.022. On the contrary, when the time shifting reaches 525 ms, both e_1 and e_2 reach the maximum (about 0.5). In addition, it also implies that the cycle for chosen walking scenarios is around 1050 ms. It is worth noting that the change of Tx power may also lead to the variation of e .

B. Simulation Results

1) **Throughput:** Fig. 18 shows the average total throughput of the proposed A3NC scheme in comparison with the Con-TDMA, Only-AA and TDMA+DNC schemes in four typical walking scenarios, with consideration of four different deployment directions on the wrists. The Tx power and β_r are set at -38 dBm and 0.9 respectively. The “Total Throughput” here means

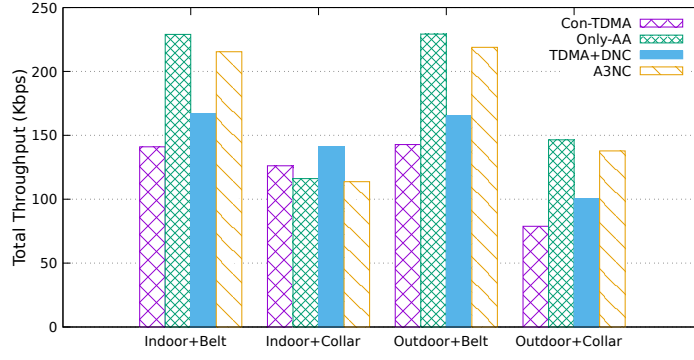


Fig. 18. Throughputs of typical walking scenarios.

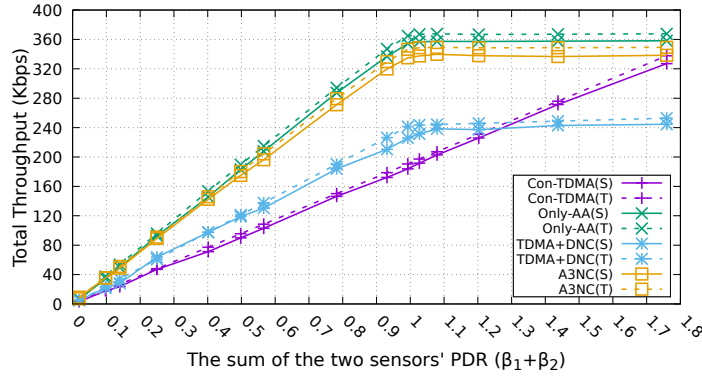


Fig. 19. Throughput vs. $\beta_1 + \beta_2$.

the sum of the average upload throughput of the two sensors.

As shown in Fig. 18, the performances of Only-AA and A3NC are slightly below that of Con-TDMA and TDMA+DNC in the “Indoor+Collar” scenario. The reason is that, compared to the cases where the hub is deployed in the belt, the PLD is relatively small when the hub is located on the back collar. Furthermore, the significant reflection effect in the indoor environment offsets the PLD between the two channels. Nevertheless, we find that Only-AA and A3NC achieve a significant improvement in terms of upload throughput in the other three scenarios, which confirms the feasibility of AA method in the real walking scenarios.

Next, we explore relationship between the upload throughput and three key parameters: $\beta_1 + \beta_2$, β_r and e . Firstly, Fig. 19 shows the results of total throughput vs. $\beta_1 + \beta_2$, when $\beta_r = 0.9$ and the time shifting is 0 ms. The Tx power of two sensors is amplified to increase the sum of β_1 and β_2 , hence the value of e is also augmented. The dash lines correspond to theoretical results derived from mathematical analyses in Section V. The solid lines indicate the simulation results. Fig. 19 clearly shows the good agreement between the mathematical analyses in Section V and the simulation results. The small deviation between the analytical and simulation curves is mainly due to two reasons. Firstly, the theoretical formulas analyze the throughput at the bit level, but the simulation model works at the packet level. The overhead of the packet header is not taken into account in the theoretical analyses. Meanwhile, because of the packetization of data, the wastage of the tail end of *SUI* (when the duration is too short to complete the transmission of one packet) is inevitable. Secondly, the idle interval between the transmission of two packets is not considered in the theoretical analysis.

Both theoretical and simulation results prove that Only-AA and A3NC achieve a remarkable improvement in comparison

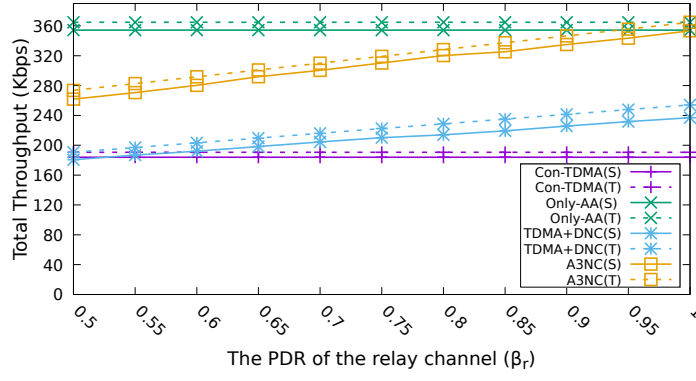


Fig. 20. Throughput vs. β_r .

with Con-TDMA and TDMA-DNC, especially when $\beta_1 + \beta_2 = 1$. Also, it can be seen that the performances of Only-AA, TDMA+DNC and A3NC plateau when the $\beta_1 + \beta_2 > 1$. For the two schemes with AA mechanism, based on the Eqs. (12) and (21), the key part determining the throughput of Only-AA and A3NC is $\beta_1 + \beta_2 - 2e$, i.e., $(\beta_1 - e) + (\beta_2 - e)$. However, since the signal collisions increase with the amplifying of Tx power, $\beta_1 - e$ and $\beta_2 - e$ are limited under 0.5. Therefore, after $\beta_1 + \beta_2$ reaching one, further increase of β_1 or β_2 cannot improve the total throughput. For the TDMA+DNC scheme, based on the analysis result in Eq. (18), if $\beta_r = 0.9$ and $\beta_1 + \beta_2 > 1$, the increment of β_1 or β_2 affects the throughput slightly. The reason is that, if the probability of successful relay cooperation is close to one, the reception of either one packet from the two sensors can lead to the decoding of the other packet. In other words, when $\beta_1 + \beta_2 \geq 1$, the hub can receive both two sensors' packets which is the best situation in one transmission process. Therefore, the upload throughput of the TDMA+DNC is saturated if $\beta_1 + \beta_2 > 1$. As for the Con-TDMA, the *SUI* of the two sensors are separated in the time domain and the relay node is not adopted, so the performance of Con-TDMA scheme raises linearly with the increase of $\beta_1 + \beta_2$.

Fig. 20 shows the throughput performance as a function of β_r , with the Tx power of -38 dBm and the time shifting of 0 ms. As expected, the performances of both TDMA+DNC and A3NC improve with the increase of β_r , while those of Only-AA and Con-TDMA remain constant. Besides, when the time shifting is 0 and Tx power is -38 dBm (i.e., $\beta_1 + \beta_2 < 1$), e is small enough to be ignored (around 0.022). Therefore, based on Eq. (18) and Eq. (21), the slope of TDMA+DNC is $1/3(1 + \beta_r)$, while that slope of A3NC is $1/2(1 + \beta_r)$. That is the reason that the curve of A3NC is steeper than that of TDMA+DNC. In other words, A3NC is more sensitive to the relay channel condition, which is acceptable when the relay node is deployed outside the human body.

In Figs. 19 and 20, the time shifting is set to 0 ms, which also matches the normal swing motion of arms in walking scenarios. However, if we increase the time shifting, the resulting change of e will affect the system performance. Fig. 21 shows the throughput performance as a function of e , when the other parameters are set to the default value (including $\beta_1 + \beta_2 = 0.99$, and $\beta_r = 0.9$). As expected, the performance of Con-TDMA is not affected by e , and the performances of the two schemes with AA mechanism decrease with the increase of e . Moreover, since $\beta_r = 0.9$, the slope of two AA schemes are almost the same. An exception is that, although AA mechanism is not adopted in TDMA+DNC, its performance still declines with the rise of e , which does not match the analysis result in Eq. (18). The reason for this exception is that due to the two *SUIs* of the two sensors are adjacent and are relatively short compared to the walking cycle, when e increases by changing the swing

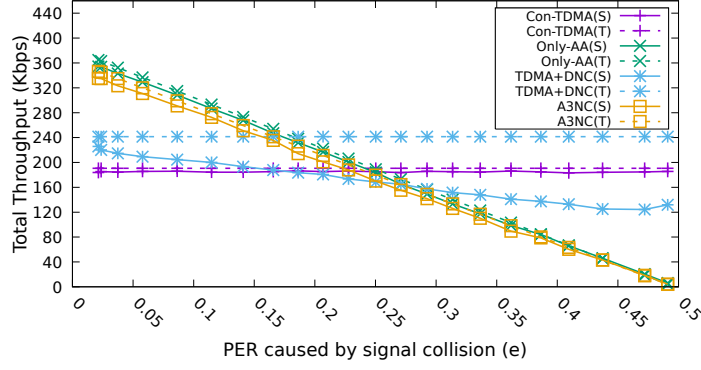


Fig. 21. Throughput vs. e .

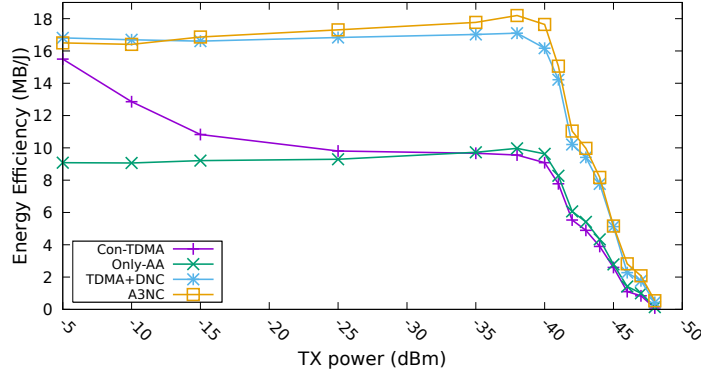


Fig. 22. Measured energy efficiency (MByte/Joule) vs. Tx power.

behavior of two arms, the hub tends to receive or discard both packets from the two sensors. Accordingly, the effectiveness of the relay cooperation is reduced.

2) **Energy Efficiency:** Fig. 22 compares the energy efficiency of the four schemes when β_r and time shifting are set to 0.9 and 0 ms respectively. Due to both the PDR sum of the two sensors (i.e., $\beta_1 + \beta_2$) and the energy consumption are affected by the Tx power of the sensors, we take the Tx power to evaluate the energy efficiency. The “Energy Efficiency” (EE) in Fig. 22 represents how many megabytes (MB) of data are transmitted by consuming one joule energy of the two sensors (cf. Eq. (22)). As mentioned before, the energy of the relays and hub are not considered in this paper.

From Fig. 22, the energy efficiency of Only-AA, A3NC and TDMA+DNC schemes increase slightly when the Tx power decreases from -5 dBm to -38 dBm, and reduces rapidly when Tx power is below -38 dBm. The curve trend can be explained by two reasons. First, combining the simulation results in Fig. 19 and Fig. 16, when the Tx power is below -38 dBm (i.e., $\beta_1 + \beta_2 < 1$), the upload throughputs rise up with the increase of the Tx power, and the throughputs reach a plateau after Tx power ≥ -38 dBm. On the other hand, the working power of the sensor slightly reduces with the decline of TX power. Accordingly, based on the definition of EE in Eq. (22), the energy efficiency reaches the maximum at the point of -38 dBm. As for the Con-TDMA scheme, since its throughput declines continuously with the decrease of Tx power (see Fig. 19), its energy efficiency experiences a similar trend. At last, combining Fig. 19 and Fig. 22, when the Tx power is -38 dBm, both the throughput and the energy efficiency reach the maximum. The Tx power -38 dBm is considered as the optimal value for Only-AA, TDMA+DNC and A3NC for the simulation configuration.

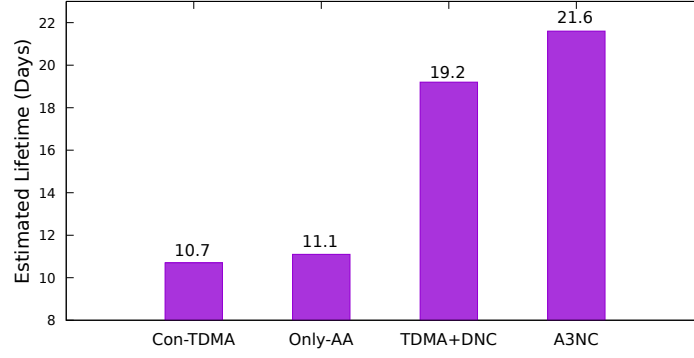


Fig. 23. The comparison of the estimated lifetime.

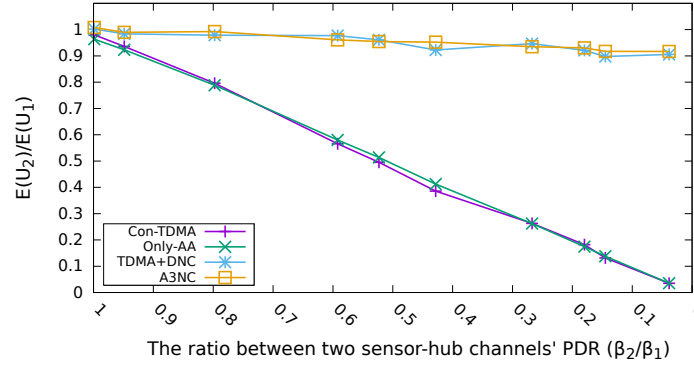


Fig. 24. The comparison of throughput between the two sensors.

We also estimate the lifetime for a typical WBAN application, in which an upload traffic of 200 kbps is required for each sensor equipped with one CR2032 button battery. In this case, the TX powers of the two sensors are set to the optimal value: -38 dBm. Fig. 23 shows that the estimated lifetime of Con-TDMA or Only-AA is less than 12 days, while A3NC lasts over 21 days, which is a significant enhancement for WBANs systems.

3) **Throughput Balance:** All the above results are derived when the PDRs of two sensors are approximately equal, i.e., $\beta_1 \approx \beta_2$. Here we consider the case where β_1 and β_2 are different.

Fig. 24 shows the average ratio between the upload throughput of the two sensors, $E(U_2)/E(U_1)$, as a function of β_2/β_1 . The curves of both TDMA+DNC and A3NC are almost constant and are approximately equal to one, i.e., $E(U_1) \approx E(U_2)$. That means the two schemes with relay cooperation can achieve a better throughput balance between the two sensors. In contrast, when the relay node is not utilized (Con-TDMA and Only-AA schemes), the difference between $E(U_1)$ and $E(U_2)$ increases significantly with the decrease of β_2/β_1 . The simulation results in Fig. 24 also confirm our analysis in Table IV.

In summary, the simulation results confirm the theoretical analyses in Section V. In contrast to conventional methods, the newly proposed A3NC scheme achieves a significant throughput improvement (about 10%-40% in normal walking scenarios), while attains substantial power saving. Besides, with the help of cooperation communication, A3NC maintains a good throughput balance between two correlated sensor nodes.

VII. CONCLUSION

In this paper, first, we introduced our customized portable wireless transceivers, which enable us to collect the realistic on-body wireless channel data in the walking scenarios. The statistical results of these collected datasets show that there exists a significant path loss discrepancy between two on-body channels when people are walking. This feature of on-body channel confirms the rationale of our newly proposed cross-layer network coding scheme, called A3NC. The A3NC scheme integrates the ANC technology in the PHY layer and a timeslot allocation algorithm in the MAC layer, to improve transmission performance in the walking scenarios. In A3NC, two sensors with significant path loss discrepancy simultaneously transmit packets to the relay and hub. Then, the relay forwards network coding packets to the hub and the hub performs the decoding process after the first two phases. We also derived the mathematical analyses for the proposed A3NC from three perspectives, namely sum throughput, energy efficiency and throughput balance, in comparison with three other existing schemes. The validity of these performance analyses has been confirmed by our simulation. Simulation results show that the A3NC scheme achieves the best trade-off between the energy efficiency and the throughput, compared to the other three existing schemes. Meanwhile, A3NC scheme provides a better sum throughput than the Con-TDMA and TDMA-DNC and roughly the same sum throughput as the Only-AA one. To the best of our knowledge, this is the first work attempting to bridge network coding to a spatially correlated WBAN system. For the future work, we plan to deepen the exploration of the on-body WBAN channels by considering more sensors and more generalized scenarios.

REFERENCES

- [1] S. Movassaghi *et al.*, "Wireless body area networks: A survey," *IEEE Commun. Surveys Tuts.*, vol. 16, no. 3, pp. 1658–1686, Mar 2014.
- [2] *IEEE Standard for Local and metropolitan area networks - Part 15.6: Wireless Body Area Networks*, IEEE Std. 802.15.6, 2012.
- [3] Yazdandoost, K. Yekeh, and K. Sayrafian-Pour, "Channel model for body area network (BAN)," *IEEE P802.15.6*, 2009.
- [4] J. N. Laneman, D. N. C. Tse, and G. W. Wornell, "Cooperative diversity in wireless networks: Efficient protocols and outage behavior," *IEEE Trans. Inf. Theory*, vol. 50, no. 12, pp. 3062–3080, Dec 2004.
- [5] L. C. Tran and A. Mertins, "Error performance and energy efficiency analyses of fully cooperative OFDM communication in frequency selective fading," *IET Commun.*, vol. 10, no. 18, pp. 2525–2533, Dec 2016.
- [6] S. Ivanov, D. Botvich, and S. Balasubramaniam, "Cooperative wireless sensor environments supporting body area networks," *IEEE Trans. Consum. Electron.*, vol. 58, no. 2, pp. 284–292, May 2012.
- [7] L. C. Tran *et al.*, "Comprehensive performance analysis of fully cooperative communication in WBANs," *IEEE Access*, vol. 4, pp. 8737–8756, Dec 2016.
- [8] Y. Chen *et al.*, "Cooperative communications in ultra-wideband wireless body area networks: Channel modeling and system diversity analysis," *IEEE J. Sel. Areas Commun.*, vol. 27, no. 1, pp. 5–16, Jan 2009.
- [9] R. Ahlswede *et al.*, "Network information flow," *IEEE Trans. Inf. Theory*, vol. 46, no. 4, pp. 1204–1216, Jul 2000.
- [10] S. Katti *et al.*, "Xors in the air: Practical wireless network coding," *IEEE/ACM Trans. Netw.*, vol. 16, no. 3, pp. 497–510, Jun 2008.
- [11] D. S. Lun, M. Médard, and R. Koetter, "Efficient operation of wireless packet networks using network coding," in *Proc. International Workshop on Convergent Technologies (IWCT)*, Jun 2005.
- [12] G. E. Arrobo and R. D. Gitlin, "New approaches to reliable wireless body area networks," in *Proc. IEEE International Conference on Microwaves, Communications, Antennas and Electronic Systems (COMCAS 2011)*, Nov 2011, pp. 1–6.
- [13] G. E. Arrobo and R. D. Gitlin, "Improving the reliability of wireless body area networks," in *Proc. Annual International Conference of the IEEE Engineering in Medicine and Biology Society*, Aug 2011, pp. 2192–2195.
- [14] P. M. P. Dharshini and M. Tamilarasi, "Adaptive reliable cooperative data transmission technique for wireless body area network," in *Proc. International Conference on Information Communication and Embedded Systems (ICICES2014)*, Feb 2014, pp. 1–4.

- [15] E. Kartsakli *et al.*, “A cloud-assisted random linear network coding medium access control protocol for healthcare applications,” *Sensors*, vol. 14, no. 3, pp. 4806–4830, Mar 2014.
- [16] Z. J. Haas and T. C. Chen, “Cluster-based cooperative communication with network coding in wireless networks,” in *Proc. IEEE Military Communications Conference*, Oct 2010, pp. 2082–2089.
- [17] X. Shi, M. Médard, and D. E. Lucani, “When both transmitting and receiving energies matter: An application of network coding in wireless body area networks,” in *Proc. International Conference on Research in Networking*. Springer, May 2011, pp. 119–128.
- [18] S. Movassaghi, M. Shirvanimoghaddam, and M. Abolhasan, “A cooperative network coding approach to reliable wireless body area networks with demodulate-and-forward,” in *Proc. 9th International Wireless Communications and Mobile Computing Conference (IWCMC)*, Jul 2013, pp. 394–399.
- [19] H. Alshaheen and H. T. Rizk, “Improving the energy efficiency for biosensor nodes in the WBSN bottleneck zone based on a random linear network coding,” in *Proc. 11th International Symposium on Medical Information and Communication Technology (ISMICT)*, Feb 2017, pp. 59–63.
- [20] A. Taparugssanagorn, F. Ono, and R. Kohno, “Network coding for non-invasive wireless body area networks,” in *Proc. 20th IEEE International Symposium on Personal, Indoor and Mobile Radio Communications Workshops (PIMRC Workshops)*, Sept 2010, pp. 134–138.
- [21] M. A. Razzaque *et al.*, “QoS-Aware error recovery in wireless body sensor networks using adaptive network coding,” *Sensors*, vol. 15, no. 1, pp. 440–464, 2014.
- [22] E. Byrne and A. Manada, “On coding schemes for wireless body area networks,” in *Proc. International Symposium on Communications and Information Technologies (ISCIT)*, Oct 2012, pp. 149–154.
- [23] S. Movassaghi *et al.*, “An energy efficient network coding approach for wireless body area networks,” in *Proc. 38th IEEE Conference on Local Computer Networks (LCN)*, Oct 2013, pp. 468–475.
- [24] E. Kartsakli *et al.*, “Reliable MAC design for ambient assisted living: moving the coordination to the cloud,” *IEEE Commun. Mag.*, vol. 53, no. 1, pp. 78–86, Jan 2015.
- [25] A. S. Lalos *et al.*, “RLNC-Aided cooperative compressed sensing for energy efficient vital signal telemonitoring,” *IEEE Trans. Wireless Commun.*, vol. PP, no. 99, pp. 1–1, Jul 2015.
- [26] R. D’Errico and L. Ouvry, “Time-variant BAN channel characterization,” in *Proc. 20th IEEE International Symposium on Personal, Indoor and Mobile Radio Communications (PIMRC)*, Sept 2009, pp. 3000–3004.
- [27] H. Feng *et al.*, “Prediction-based dynamic relay transmission scheme for wireless body area networks,” in *Proc. 24th IEEE International Symposium on Personal, Indoor, and Mobile Radio Communications (PIMRC)*, Sept 2013, pp. 2539–2544.
- [28] H. Zhang, F. Safaei, and L. C. Tran, “Joint analog network coding and channel allocation in the walking scenario for WBAN,” in *Proc. IEEE Symposium on Computers and Communication (ISCC)*, Jun 2016, pp. 604–609.
- [29] H. Zhang, F. Safaei, and L. C. Tran, “Measurement-based characterizations of on-body channel in the human walking scenario,” *to appear in Proc. 85th IEEE Vehicular Technology Conference (VTC2017-Spring)*, Jun 2017.
- [30] L. Hanlen *et al.*, “Open-source testbed for body area networks: 200 sample/sec, 12 hrs continuous measurement,” in *Proc. 21st IEEE International Symposium on Personal, Indoor and Mobile Radio Communications Workshops*, Sept 2010, pp. 66–71.
- [31] M. Lauzier *et al.*, “Full mesh channel measurements on body area networks under walking scenarios,” in *Proc. 7th European Conference on Antennas and Propagation (EuCAP)*, Apr 2013, pp. 3508–3512.
- [32] K. Pearson, “Note on regression and inheritance in the case of two parents,” *Proceedings of the Royal Society of London*, vol. 58, pp. 240–242, Jan 1895.
- [33] J. G. Proakis and M. Salehi, *Digital Communications*, 5th ed. McGraw-Hill, 2007.
- [34] J. G. Proakis and M. Salehi, *Fundamentals of communication systems*. Pearson Education, 2007.
- [35] S. Katti, S. Gollakota, and D. Katabi, “Embracing wireless interference: Analog network coding,” *SIGCOMM Comput. Commun. Rev.*, vol. 37, no. 4, pp. 397–408, Aug 2007.
- [36] Castalia: a wireless sensor network simulator. [Online]. Available: <https://github.com/boulis/Castalia/>