An Efficient Polar Coding Scheme for Uplink Data Transmission in Narrowband Internet of Things Systems

RABIU SALE ZAKARIYYA 1, (Graduate Student Member, IEEE), KHALID HOSSEN JEWEL 1, AKINWALE O. FADAMIRO 1, (Member, IEEE), OLUWOLE JOHN FAMORIJI 2, AND FUJIANG LIN 1, (Senior Member, IEEE)

1Micro/Nano-Electronic System Integration Research and Development Centre (MESIC), University of Science and Technology of China (USTC), Hefei 230026, China
2Institute of Marine Electronic and Intelligent System, Zhejiang University at Zhoushan Campus, Zhoushan 316021, China

Corresponding author: Rabiu S. Zakariyya (rabiu123@mail.ustc.edu.cn)

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ABSTRACT
Narrowband Internet of Things (NB-IoT) is an emerging IoT cellular-based wireless technology that builds based on the existing LTE standard to extend its network coverage suited for Low Power Wide Area Network (LPWAN) connectivity. In the NB-IoT system, LTE turbo codes and Cyclic Redundancy Check (CRC) codes are adopted as the main channel coding technique for Narrowband Physical Uplink Shared Channel (NPUSCH). However, turbo codes require iterative decoding algorithm at the expense of high computation complexity of decoding hence is crucial to achieving the LPWAN requirement of NB-IoT systems. Polar codes are best suited in this regard due to its capacity approaching and complexity reduction in short block length code. In this paper, we proposed an efficient polar coding technique using the Belief Propagation (BP) decoding algorithm for uplink data transmission on the NPUSCH channel. Furthermore, the BP algorithm is incorporated with the CRC stoppage criterion to decrease the number of decoding iteration and reduce computational complexity. In this scheme, single-tone numerology of NPUSCH using 3.75 kHz and 15 kHz subcarrier spacing is adopted. Then, the encoded data is generated with different NPUSCH resources. The theoretical formulation and simulation demonstrate that the proposed scheme provides better error rate performance over the adopted LTE turbo codes and other polar decoding algorithms while reducing the computational complexity.

INDEX TERMS
Narrowband IoT (NB-IoT), BP polar decoding algorithm, narrowband physical uplink shared channel (NPUSCH), cyclic redundancy check (CRC), block error rate (BLER).

I. INTRODUCTION
In the recent decade of wireless connectivity, the Internet of things (IoT) technology has recently become the modern communication technology due to its involvement in smart connected applications such as smart cities, e-health, precision agriculture, and much more [1]–[3]. However, rapid demand IoT devices have grown in wireless technology requiring efficient transmission for low power consumption, low cost, and low complexity in terms of implementation. In 20’s, it is forecasted that over 100 billion connected devices will be enabled thereby generating huge amount of revenue [2], [4], [5]. To meet this demand, the 3rd Generation Partnership Project (3GPP) has set another communication standard specifically for IoT as part of Release 13 (Rel-13) called the Narrowband IoT (NB-IoT) [6] which regarded as the key enablers in Fifth Generation (5G) technology that designed to offer IoT service within which Long Term Evolution (LTE) and WLAN are failed. NB-IoT is a prominent wireless communication technology that co-exists within LTE network to transmit the little amount of data for IoT devices under Low Power Wide Area (LPWA) network technology [7]. To optimize NB-IoT LPWAN specification, additional features are included in both Release-14 and 15 [2] of 3GPP standard. In Release-14 features applicable for massive machine-type communication (mMTC) use
case which includes device positioning/localization, and cell search enhancement are introduced focuses on high peak data rates with UE power allocation of 23 dBm and enhanced cell ID. Release-14 [8] integrates NB-IoT into 5G including low latency reduction, new wake up receiver design for efficient operation, higher density support, and multiple Hybrid Automatic Repeat Request (HARQ) feedback which achieve a better throughput with repetitive transmission. It is expected to achieve long-lasting of battery life (least 10 years), enhanced coverage (more than Bluetooth and GPRS), low latency (less than 10 msec), massive connection (about 10,000 devices per cell), low cost (less than $1 per module), and the peak data rate of about 250 kbps per second [9].

Considering these new specifications in NB-IoT systems, algorithms such as uplink data transmission, and channel coding are redesigned to optimize uplink transmission in NB-IoT system. However, for data transmission in uplink, channel coding is based on LTE Physical (PHY) Layer forward error correction (FEC) technique [10], [11] including turbo coding and Cyclic Redundancy Check (CRC) codes are being utilized to encode and decode the user transport block (TB) on the physical channels for reliable data delivery. Each NB-IoT UE will schedule with TB data which is mapped to specific layers on uplink channels in every transmission time interval (TTI) of which allow eNodeB to decode the transmitted packet correctly [11]–[13]. With efficient channel coding scheme, better throughput will be achieved. Similarly, in 5G network, NB-IoT will require the use of qualitative channel coding is the key to provide global coverage and massive IoT connection, as proposed in 3GPP Releases 14 and 15, respectively. Unfortunately, the use of turbo codes in NB-IoT does not guarantee a low complexity support as well as good error rate performance for short block length, especially in the error floor region. Therefore, an efficient channel coding algorithm need to be employed to the data transmission specifically in the data transmission channel for reliable transmission that can be suited for NB-IoT LPWA requirements.

The polar code is among the modern FEC coding technique that gains a lot of recognition in recent communication standards, and recommends by 3GPP standard in Release 15 as the PHY layer error correction technique for the enhanced mobile broadband (eMBB) to replaces the tail-biting convolutional codes (TBCCs) in LTE standard [14], [15]. However, recent research work considers polar codes as a candidate for mMTC, URLLC, and implemented into an ASIC chip hardware [16], [17]. Polar codes were initially proposed by Arikan [18] based on the concept known as the channel polarization that theoretically achieved the capacity limit of the discrete-input memoryless channels (DMCs) with explicit construction approach. The decoding of polar codes are based on the successive cancellation (SC) [19]–[22] decoding algorithm which exhibits low complexity implementation but has a poor performance against LDPC and turbo codes, but still improving with other decoding algorithms. Similarly, the Belief propagation (BP) is another useful algorithm that improve the polar decoding performance [17], [23], [24]. One advantage of the BP algorithm is its highly parallel implementation with superior performance over the SC and SCL decoding algorithms.

To utilize the advantage of short information bits in terms of low power consumption, low decoding complexity, and low hardware resource utilization, Belief Propagation (BP) of polar codes is the best candidate suited for that. On this basis, we are considering the polar coding technique for uplink data transmission in NB-IoT systems. We proposed a BP decoder concatenated with CRC (BPC) error detection technique to further reduce the decoding complexity for different decoder block size so that each coded data on NPUSCH can be transmitted to the extent of a high degree of reliability without the need for retransmission. The main contributions of this paper are itemized as follows.

- We first formulate a system model of uplink NB-IoT with multiple IoT devices communicate with a base station (BS), including the PHY channel coding, SC-FDMA modulation, and layer mapping.
- We then introduce the basic principle of the polar coding scheme to the NPUSCH channel as the channel coding for efficient data transmission. We utilize the polar BP decoding algorithm to improve uplink transmission reliability. To further reduce decoding complexity, we employ CRC early stoppage mechanism in our decoding algorithm so that the number of decoding iteration can be controlled.
- Finally, we perform both Monte Carlo and a link-level simulation to obtain the performance of our algorithm in terms of Block Error Rate (BLER) performance compared with different polar decoding algorithms. Simulation results show to justify the effectiveness our proposed scheme, which significantly improves the transmission reliability of the NPUSCH channel.

The remainder of this paper is organized as follows. In section II, we present the related work for uplink NB-IoT system. The general overview of data transmission in the uplink NB-IoT system will be presented in section III. The uplink PHY layer transmission functionalities are outlined in section IV. In section V, we present the BPC based polar decoding algorithm and its analytical expressions. In section VI, we present numerical simulations, and finally conclusion remarks in section VII.

II. RELATED WORK

Recently, with the emergence of 5G communication technology, the NB-IoT presents multiple opportunities in both consumer and industrial applications, which makes it one of the hot research topics in recent years. Besides, due to specific technical requirements of LPWAN in NB-IoT, the design of uplink technology is more challenging than the existing LTE uplink systems, and hence it requires different PHY layer configuration which needs further investigations.

Authors in [25] studied the data transmission procedure to reduce the signal overhead. Also, they proposed the performance optimization, the cell capacity gain, and the UE battery...
enhancement for different coverage levels. Oh and Shin [26] proposed an efficient small data transmission scheme using a control plane (CP) method, where a small data packet can be transmitted without transmission losses and signaling overhead specifically in idle state. For efficient uplink scheduling, Yu et al. [27] investigates the NB-IoT uplink scheduling and link adaptation for data transmission by utilizing the repetition factor of NPUSCH and modulation and coding scheme (MCS) index to improve the uplink transmission performance. However, Hwang et al. [28] developed an efficient detection and synchronization algorithm to mitigate detection problem in uplink narrowband physical random access channel (NPRACH). Authors in [29] proposed NB-IoT uplink scheduling technique compatible with low earth orbit satellite (LEO) through reducing the differential Doppler down suitable to IoT devices thereby enhancing the coverage performance.

In addition, uplink NB-IoT suffers from interference and a data rate loss due to lack of orthogonality between the to the LTE signals and NB-IoT resulting performance degradation in the BS. To cope with these problems, authors in [30] proposed the channel equalization and interference under the in-band and guard-band deployment scenario. Although the proposed method is analyzed without considering the mandatory single-tone transmission which is unique from LTE systems. To improve on that, Beyene et al. [31] investigated the frequency domain channel estimation and equalization technique based on the least square (LS) estimation approach. Author in [32] studied the uplink receiver algorithm for both NPUSCH and NPRACH channels in uplink NB-IoT systems, while [33], [34] studied the data scheduling and estimation approach for NPUSCH, respectively. Authors in [35] proposed a joint channel estimation and repetition combining method to obtain better estimation precision over the conventional LS approach. In the same way, authors in [36] propose an efficient receiver design which integrates the channel estimation and the interpolation.

In our previous work [37], we presented the basic channel coding scheme for uplink NB-IoT based on the LTE turbo coding scheme using the requirement from the 3GPP standard. The works in [38] and [39] have specified the key requirements in developing qualitative channel coding techniques for NB-IoT in achieving the LPWAN design goals. Migabo et al. [40] study the energy-efficient channel coding scheme based on adaptive MCS scheme. Authors in [41] studied the CRC aided hybrid decoding for turbo codes. Yang et al. [42] proposed an energy-efficient HARQ transmission using field test analysis.

III. OVERVIEW OF UPLINK TRANSMISSION IN NB-IOT SYSTEMS

This section briefly discusses the main components related to uplink data transmission in Narrowband Physical Uplink Shared Channel (NPUSCH).

In the uplink NB-IoT system, the basic transmission technique of the uplink is similar to that of the LTE system which is built on the basis of Single Carrier Frequency Division Multiple Access (SC-FDMA) as the main multiplexing scheme, which helps in reducing complexity and low Peak-to-Average Power Ratio (PAPR) as compared to conventional OFDMA as provided in 3GPP specification [11]. The 3GPP standard introduces new concept data repetition on the new uplink physical channels in NB-IoT to extend the indoor/outdoor network coverage of LTE systems in a remote area. For this purpose, a flexible transmission scheme has been recognized as a key leading towards the LPWAN requirement of NB-IoT that enables a number of NB-IoT users to simultaneously transmits data on the physical channels to the destination for network access. However, the operating bandwidth required one piece of LTE Physical Resource Block (PRB) corresponds to 12 subcarriers with each of 15 kHz making a total of 180 kHz as in the downlink transmission respectively. The transmission mechanism is organized in a form of frames which are shared among the physical uplink channels. With this adoption, new physical channels are defined consisting of the fraction of preambles and user data which are allocated on a dedicated uplink physical channels. These are the NPRACH for uplink random access preamble and NPUSCH for both data and control transmission which allows NB-IoT users to upload their data successfully to the BS over a wireless channel.

For NPUSCH data transmission, transmission is carried out on a specific narrowband subcarrier depending on the choice of subcarrier spacing [13], i.e., single-tone or multi-tone transmission. Multi-tone transmission is known to be optional numerology which has 3, 6, 12 as an acceptable number of tones using the same LTE 15 kHz SCS. However, for single-tone transmission, 3.75 kHz and 15 kHz are the acceptable SCS option. For multi-tone transmission, the RU scheduling assignment composed of 1 ms for 12 tones, 2 ms for 6 tones, 4 ms for 3 tones, 8 ms for a single-tone, 4 ms for 3 tones, 2 ms for 6 tones, and 1 ms for 12 tones. While for single-tone transmission, 32 ms for 3.75 kHz. The whole frame scheduling is illustrated in Figure 1. Note that the 15 kHz SCS is similar to LTE and provides better performance.

When adopting single-tone transmission, new radio frame structures are defined in a time-domain perspective.
The whole frame composed of 10 identical subframes. Each subframe is segmented into two equal slots of 0.5 ms each. Each slot represents 7 OFDM symbols, and one of the symbols is allocated to DMRS (Demodulation Reference Signals) to estimate the NPUSCH symbols as illustrated in Figure 2.

### IV. POLAR CODED NB-IoT UPLINK PHY LAYER MODEL

This section will present the uplink PHY layer model of NB-IoT with NPUSCH transmission, including the including the PHY layer CRC encoding, polar coding scheme, uplink subcarrier mapping using NPUSCH single-tone transmission, and other protocol related to uplink data transmission in NB-IoT.

#### A. POLAR CODING SCHEME

Let us consider the uplink PHY network of NB-IoT system, where multiple of IoT UEs indexed by $k = \{1, \ldots, K\}$ simultaneously requesting uplink communication to a single base station or eNodeB. The $k$-th UE can simultaneously transmit identical TBS on their NPUSCH channel to the eNodeB for network access. Each UE has a certain amount of TB to be transmitted to the eNodeB for network access, and each UE is linked with different MCS index and the number of RU selected from the available uplink TBs listed in Table 1.

Before NPUSCH encoding, at first, its sequence of information bits stream of $u \in \{u_1, u_2, \ldots, u_N\}$ in a form of TBS data of length $N$ are forwarded to MAC layer for error detection purpose. The $u$ TB are concatenated with CRC redundant parity bits $u(x) = u_0x^0 + u_1x^1 + \cdots + u_{m-1}x^{m-1}$ for CRC encoding. The CRC code are represented by its generator polynomial $g(x) = g_0x^0 + \cdots + g_{n-m-1}x^{n-m-1} + x^{n-m}$, where the computation are perform over Galois field $GF(2)$ arithmetic. In CRC encoding, the information bits message are free multiply $x^{n-m}$ followed by the division of $x^{n-m} \cdot u(x)$ by $g(x)$ which result to a check bits known as reminder $e(x)$.

Thus, the encoded CRC bits are related as

$$\bar{u}(x) = \bar{q}(x) \cdot g(x) + e(x),$$

where $\bar{q}(x)$ is the quotient over $GF(2)$. The presence of an error in $\bar{u}(x)$ message can be detected whenever $e(x) \neq 0$ polynomial, otherwise $\bar{u}(x)$ is free from error. The output of CRC encoder $\bar{u}(x)$ are fed to polar encoder to perform error correction coding. To encode the NPUSCH data using polar encoding, we consider the construction of $N = 2^m$ polar code represented by $(N, M, A, C_{Av})$, has a corresponding code rate $R = M/N$. Where $A$ denotes the indices of $\bar{u}(x)$ bits, and $C_{Av}$ is the frozen bits of the remaining codeword. Then the signals $\bar{u}(x)$ are transformed into a polar code codeword sequence $v_k$ given by

$$v_k = w_k G_N = w_k B_N F_{m}^\otimes,$$

(2)

$$F = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix},$$

(3)

where $F_{m}^\otimes$ denotes the $m$-th Kronecker power of polarization kernel and $B_N$ is the permutation matrix define in [23]. The encoded polar sequence $v_k$ is then mapped into BPSK constellation to produced the complex modulated vector $X$ for $k = 0, \ldots, N-1$, expressed as

$$x_k = [x_0, x_1, \cdots, x_{N-1}]^T,$$

(4)

The resultant vector are then processed by SC-FDMA modulator for uplink subcarrier mapping.

#### B. UPLINK TRANSMISSION SCHEME

The simplified PHY layer configuration of uplink NB-IoT system model is illustrated in Figure 3. It consists the basic structure of the $k$-th UE SC-FDMA modulation on the upper part and the single BS of eNodeB on the lower part of Figure 3 respectively. Let us first consider the analysis of $k$-th UE SC-FDMA transmission model in the transmitter side. The NPUSCH coded signal sequence $x_k$ is precoded from time to frequency domain transformation through $N_v$-point discrete Fourier transform (DFT) the along $n$-th subcarrier is

![FIGURE 2. The uplink resource grid structure for NPUSCH.](image-url)
represented as follows

\[ x_n = \frac{1}{\sqrt{N_b}} \sum_{k=0}^{N_b-1} x_k e^{-j2\pi kn/N_b}, \quad 0 \leq k \leq N_b - 1, \quad (5) \]

Since we are using multi-user SC-FDMA scheme, a localized subcarrier mapping process is employed in our model as in LTE subcarrier mapping [36]. As a result, adjacent subcarrier can be allocated among each of the \( k \)-th users in time domain. Each user carried out data transmission on multiple number of RUs, each RU contain number of slots depending on the choice of subcarrier spacing (SCS). Note that in this paper we only consider the single-tone transmission with both 3.75 kHz and 15 kHz SCS respectively. Then, the scheduled elements in \( x_m \) are converted to time domain through \( M \)-point inverse DFT (IDFT) operation resulting to \( x_m \) along \( m \)th subcarrier as

\[ x_m = \frac{1}{\sqrt{M}} \sum_{n=0}^{M-1} x_n e^{-j\frac{2\pi mk}{M}}, \quad 0 \leq k \leq M - 1. \quad (6) \]

A Cyclic Prefix (CP) sequence is appended on the IDFT codeword to block inter-symbol interference (ISI) occurrence which is then propagated through the wireless channel. On the receiving side (eNodeB), the CP is dropped, by performing the localized subcarrier demapping, the received signal sequence are processed through the \( N_t \)-point DFT (IDFT) operation resulting to \( x_k \) along \( n \)th subcarrier as

\[ Y(n) = \sum_{n=0}^{N_t-1} y(n)e^{-j2\pi nk/N_t}, \quad 0 \leq k \leq N_t - 1, \quad (7) \]

where \( y(n) \) is the received samples after CP is removed. Simplifying further, the signal is then processed by an \( M \)-point inverse DFT (IDFT) after subcarrier inverse -mapping and without loss of generality, we assume that the channel are and free from error. Then, the equivalent time domain received signal is given packet \( Y(k) \) is related as

\[ Y(k) = X(k)H(k) + W(k), \quad (8) \]

and

\[ H(k) = \text{diag} (h_{p,1}, h_{p,2} \cdots, h_{p,k}) \quad (9) \]

where \( H(k) \) is the channel gain matrix between the \( k \)-th UE and eNodeB, \( h_{p,k} \) denotes the diagonal elements of \( H(k) \) in the \( p \)-th block and \( W(k) \) is the additive white Gaussian noise (AWGN) with zero mean, variance \( \sigma \), and spectral power density \( N_0 \). The signal \( Y(k) \) is soft demodulated through BPSK demodulator, whose soft demodulated vector given

\[ \land (u_k) = \land (u_0), \land (u_1), \cdots, \land (u_N), \quad (10) \]

Based on the demodulated vector, the generated log-likelihood ratios (LLRs) of the coded bits \( u_k \) given by

\[ L(\hat{u}_k) = \log \frac{\text{prob} (u_k = 1 | x_k)}{\text{prob} (u_k = 0 | x_k)}. \quad (11) \]

Finally, the soft demodulated coded bits of \( L(\hat{u}_k) \) are decoded using the proposed polar decoding algorithm we will be our main focus in the next section.

V. DECODING ALGORITHM

In this section, we present the decoding method of NPUSCH data using the BP decoding algorithm incorporated with CRC aided scheme which is specifically design for uplink data transmission in NB-IoT systems. We set a threshold to update the computation of each iteration.

A. BP POLAR DECODING ALGORITHM

The basic decoding process of BP algorithm for polar codes are based on the factor graph illustrated in Figure 4. To elaborate further, let us consider the decoding process illustrated in Figure 4 (b), for a given \( d \) number of iteration, the decoding message sequence are updated iteratively from the left-most nodes \( L_{i,j} \) then to the right-most nodes \( R_{i,j} \). The computations of each message exchange within the factor graph expressed in Eq. (12) as follows

\[ L_{i,j}^d = f \left( L_{i,j}^{d-1} + L_{i+N/2,j+1}^{d-1} + R_{i+N/2,j}^d \right) \]

\[ L_{i+N/2,j}^d = f \left( L_{i,j}^{d-1} + R_{i,j}^{d-1} + L_{i+N/2,j+1}^{d-1} \right) \]

\[ R_{i,j}^d = f \left( R_{i,j}^{d-1} + L_{i+N/2,j+1}^{d-1} + R_{i+N/2,j}^d \right) \]

\[ R_{i+N/2,j+1}^d = f \left( R_{i,j}^d + L_{i,j+1}^{d-1} + R_{i+N/2,j}^d \right) \quad (12) \]
where \( f(a, b) = \text{sign}(a) \text{sign}(b) \min(|a|, |b|) \), which can be simplified using min-sum approximation [29] written as

\[
f(a, b) = \alpha \times \text{sign}(a) \text{sign}(b) \min(|a|, |b|),
\]

where \( \alpha \) is a scaling factor chosen around 0.9375 which can be used to improve decoder performance.

Initially, from Figure 4 (b), the rightmost nodes are initialized with 0 and \( \infty \) based on the indices of both received bits and frozen bits as Eq. (14), and the left-most nodes of the received LLR bits as Eq. (15).

\[
R_{0,i}^d = \begin{cases} 0, & \text{if } i \in A \\ \infty, & \text{if } i \in A^C \end{cases}
\]

(14)

\[
L_{L,i}^0 = \ln \frac{P(y_i | x_i = 0)}{P(y_i | x_i = 1)},
\]

(15)

B. PROPOSED DECODING ALGORITHM

To improve on our BP decoding algorithm, we introduce the CRC aided (BPC) termination scheme which can able to minimize the computational complexity of decoding iteration. After several decoding iterations are performed, the decoding decision at the BP decoder output sequence \( \hat{u}_k \) is determine based on the decision as

\[
\hat{u}_k = \begin{cases} 0, & L_{1,i} \geq 0 \\ 1, & L_{1,i} < 0 \end{cases}
\]

(16)

For \( d \)-th decoding iteration, the \( \hat{u}_i \) codeword can pass CRC check only if and only if it satisfies the stoping loop condition (i.e., \( L(\hat{u}_i) \) is larger then predetermined threshold otherwise, the iteration process will stop. Then, proceed decoding in the left-most nodes to the right-most nodes decoder until a stopping criterion is reached or the maximum number of iteration \( d_{\text{max}} \) is completed. The overall procedure is described in Algorithm 1.

**Algorithm 1** The Proposed BPC Decoding Algorithm

**Input:** The LLR of the received vector \( y_k \), and \( N \)

**Output:** Decoded message sequence \( \hat{u}_k \)

**Initialization:** Initialize \( R_{0,i}^d \) and \( L_{L,i}^0 \) from Eq. (14) to (15)

1: for \( d = 1, 2, \ldots, d_{\text{max}} \) do
2: statements..
3: for \( (1 \leq i \leq N) \) do
4: Update \( L_{i,j}^d \) and \( L_{i+N/2,j}^d \) using (12)
5: end for
6: for \( (1 \leq j \leq j + 1) \) do
7: Update \( R_{i+1,j+1}^d \), and \( R_{i+N/2,j+1}^d \) using (12)
8: end for
9: if all the decoded codeword pass CRC checks then
10: Terminate decoding and update \( \hat{u}_k \)
11: end if
12: if all the decoded codeword cannot pass CRC checks then
13: Continue to the next iteration
14: end if
15: end for
16: return the decoded output \( \hat{u}_k \) from (16)

VI. SIMULATION RESULTS

In this section, we present the numerical computation in terms of BLER of our polar coding scheme for uplink NB-IoT systems under various NPUSCH conditions. The transmission reliability of NPUSCH is the main focus in our simulation which targets to minimize the BLER so that NB-IoT users can simultaneously upload their data on NPUSCH efficiently with a high degree of reliability. Note that throughout this simulation, we assume that \( N_{\text{Rep}} = 1 \) (number of repetition for NPUSCH). The main simulation parameters used in this experiment are given in Table 2.

A. BLER SIMULATION ANALYSIS

To validate the strength of our polar coded NPUSCH scheme, we first performed a \( 10^6 \) Monte Carlo trial for to compare the BLER performance for both polar decoding schemes and our BPC algorithm, we then perform link-level simulation for NB IoT PHY layer channel coding chain according to the 3GPP NB-IoT standard [11] by incorporating the polar codes with short block length then LTE turbo codes. Furthermore, we will compare our BPC performance against other polar
TABLE 2. Simulation Parameter for NPUSCH.

| Parameter          | Value          |
|--------------------|----------------|
| System bandwidth   | 180 kHz        |
| NPUSCH SCS         | 3.75 kHz and 15 kHz |
| FFT size           | 128            |
| Cyclic prefix length | 8              |
| Modulation         | Single-tone SC-FDMA |
| Modulation order   | BPSK           |
| Maximum TBS size   | 256            |
| CRC used           | 24 bits        |
| Channel            | AWGN channel   |
| Channel Estimation | DMRS           |
| Channel decoding   | 1/3 turbo, SC, SCL, and BP |

decoding schemes (i.e. SC, SCL, and BP) and existing LTE turbo codes adopted in NB-IoT systems.

The simulation were performed with block sizes of $N = 32$ bits, $N = 64$ bits, and $N = 128$ bits. Each coded information bits is added with 24-bits defined by generator polynomial $\overline{g}_{CRC24} = x^{24} + x^{23} + x^6 + x^5 + x + 1$ as in (1). For comparison purpose, we use the LTE turbo codes in the link level simulation which are constructed using two identical recursive systematic convolutional (RSC) encoder as in [37], which described by the polynomials $g_0(D) = 1 + D^2 + D^3$ and $g_1(D) = 1 + D + D^3$ separated with internal interleaver with a 1/3 coding rate. For decoding part, we utilize the Log-MAP decoding algorithm and the maximum number of iteration $I_{max} = 7$. The SCL decoder [22] having a block length of $N = 128$ is adopted for comparison purpose. Figure 5 (a) and (b) shows the BLER simulation results of the proposed scheme for the block length $N = 64$, and $N = 256$ for 3.75 kHz SCS, respectively. It can be noticed that the the proposed scheme gains about approximately 1.6 dB and 2.2 dB over the BP, and traditional LTE turbo decoding algorithm.

Figure 6 (a) and (b) demonstrate BLER performance comparison of BPC scheme using the block length $N = 64$, and $N = 256$ using 15 kHz SCS. The proposed scheme shows better improvement against other decoding algorithm which has a better BLER performance of 1.7dB at about $10^{-2}$ which is almost the same as the LTE turbo decoding having the same block size. In view on the above mentioned observations, we noticed that the proposed decoding algorithm can achieve better decoding performance. Furthermore, our scheme gains about 0.7dB at BLER of $10^{-3}$ using $L = 32$ compared to the SCL and LTE turbo decoding algorithms.

B. BLER PERFORMANCE WITH DIFFERENT NPUSCH RESOURCES

In this subsection we will consider the BLER computation of the proposed polar coding scheme with different NPUSCH resources. For successful transmission, We choose the TBS of NPUSCH resources. Specifically, we vary the number of TBS which correspond to different MCS level in Table 1. Such that the number of TBS are assigned as follows: TBS = 16 bits for $N_{RU} = 0$, TBS = 32 bits for $N_{RU} = 1$, TBS = 32 bits for $N_{RU} = 6$, TBS = 256 bits for $N_{RU} = 2$, TBS = 256 bits for $N_{RU} = 5$, and TBS = 256 bits for $N_{RU} = 7$.

Figure 5 and Figure 6 presents the coded NPUSCH BLER simulation results of the proposed BPC scheme using the 3.75 kHz and 15 kHz SCS respectively. Therefore the error detection and correction capability using short block polar codes can achieve a reasonable BLER performance which is acceptable for NB-IoT uplink transmission to meet its minimum data rates requirement.

Besides flexibility in NPUSCH performance, we aim to evaluate the estimated decoding complexity of our proposed scheme in terms of number of decoding iterations are performed with the variation of information block length over the SNR range.

The complexity comparisons of NB-IoT based LTE turbo codes using Log-MAP decoder [37], SC/SCL polar decoder, and our decoding scheme are demonstrated in Figure 9. And also, we can notice that the SC decoder has the least decoding complexity, but its performance is not good. While our
propose scheme has reduced complexity over the BP/SCL polar decoders and the adopted LTE turbo codes in uplink NB-IoT systems.

VII. CONCLUSION

In this paper, an efficient polar coding technique is introduced for NPUSCH data transmission to improve transmission reliability and reduced the computational complexity of decoding. We studied the NPUSCH PHY layer data transmission channel based on the 3GPP specification of NB-IoT systems. We then introduce a design metric of the BP polar decoding algorithm with CRC aided early stoppage criteria. Additionally, an effective selection criterion for choosing the right information bits for BP decoder was also been presented which are compatible with NPUSCH resources in single-tone transmission. Finally, performance analysis and results demonstration verified that the proposed scheme can significantly guarantee the transmission reliability among NB-IoT users over other polar decoding algorithms and conventional turbo codes adopted in LTE while significantly reducing the computational complexity. Thus, making it suitable for the uplink NB-IoT system and other LPWA systems associated with short code blocks length that requires low latency and reliable data transmission. Our future work...
will consider the integration of the proposed scheme with a Hybrid automatic repeat request (HARQ) for the NPUSCH uplink control channel as well as ASIC/FPGA hardware implementation.

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RABIU SALE ZAKARIYYA (Graduate Student Member, IEEE) received the B.Eng. degree in electrical engineering from the Kano University of Science and Technology (KUST), Wudil, Nigeria, in 2009, and the master’s degree in electrical and computer engineering from Meliksah University, Kayseri, Turkey, in 2014. He is currently pursuing the Ph.D. degree in electronic science and technology with the Department of Electronic Science and Technology, University of Science and Technology of China (USTC), Hefei, China. His current research interests include channel coding algorithms for 5G and NB-IoT, digital baseband processing, and low-power FPGA/ASIC design and implementations.

KHALID HOSSEN JEWEL received the bachelor’s and master’s degrees from the Department of Electrical and Electronics Engineering, Islamic University, Bangladesh, in 2004 and 2005, respectively, and the M.Phil. degree in ad hoc wireless communication, in 2015. He is currently pursuing the Ph.D. degree in electronic science and technology with the Department of Electronic Science and Technology, University of Science and Technology of China (USTC), Hefei, China. His current research interests include positioning, channel estimation, and equalization algorithms for NB-IoT systems.

AKINWALE O. FADAMIRO (Member, IEEE) received the B.Eng. and M.Eng. degrees in electrical and electronics engineering with the Federal University of Technology Akure, Nigeria, and the Ph.D. degree from the Department of Electronic Science and Technology, University of Science and Technology of China (USTC), Hefei, China. His current research interests include NB-IoT baseband processing, antenna design, optimization and calibration, RF, and wireless communication.

OLUWOLE JOHN FAMORIJI received the B.Tech. degree in electrical and electronic engineering from the Ladoke Akintola University of Technology, Ogbomoso, Nigeria, in 2009, the M.Eng. degree in communication engineering from the Federal University of Technology Akure, in 2014, and the Ph.D. degree in electromagnetic field and microwave technology from the University of Science and Technology of China (USTC), Hefei, China. He is currently a Postdoctoral Researcher with the Ocean College, Institute of Marine Electronic and Intelligent System, Zhejiang University. His current research interests include microwaves, antennas and propagation, and LPWA/NB-IoT baseband processing. He is a Reviewer of the IET Journal, IEEE SENSORS JOURNAL, and among others.

FUJANG LIN (Senior Member, IEEE) received the B.S. and M.S. degrees in electrical engineering from the University of Science and Technology of China (USTC), in 1982 and 1984, respectively, and the Ph.D. degree from the University of Kassel, Germany, in 1993. He currently works as a Full Professor and a Ph.D. Supervisor with the School of Information Science and Technology, USTC, under the Chinese 1000 Talents Program. He held the position of the Executive Director of the MESIC Center. He has authored or coauthored more than 100 scientific articles, including five patents. His main research interest includes NB-IoT PHY-layer baseband chip, modeling, and IC validation of next-generation wireless technology including devices, such as GaN, FD-SOI, and FinFET. He has been serving IEEE activities in different functions since 1995, including the Chair of the Singapore Microwave Theory and Techniques (MTT)/Antenna and Propagation (AP) Chapter, a Reviewer Board Member of the IEEE Microwave Theory and Technique Society (MTT-S), and a Technical Program Committee (TPC) Member of RFIC.