ABS+ Polar Codes: Exploiting More Linear Transforms on Adjacent Bits

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Abstract— ABS polar codes were recently proposed to speed up polarization by swapping certain pairs of adjacent bits after each layer of polar transform. In this paper, we observe that applying the Arıkan transform \( (U_i, U_{i+1}) \mapsto (U_i + U_{i+1}, U_{i+1}) \) on certain pairs of adjacent bits after each polar transform layer leads to even faster polarization. In light of this, we propose ABS+ polar codes which incorporate the Arıkan transform in addition to the swapping transform in ABS polar codes. In order to efficiently construct and decode ABS+ polar codes, we derive a new recursive relation between the joint distributions of adjacent bits through different layers of polar transforms. Simulation results over a wide range of parameters show that the CRC-aided SCL decoder of ABS+ polar codes improves upon that of ABS polar codes by 0.1 dB–0.25 dB while maintaining the same decoding time. Moreover, ABS+ polar codes improve upon standard polar codes by 0.2 dB–0.45 dB when they both use the CRC-aided SCL decoder with list size 32. The implementations of all the algorithms in this paper are available at https://github.com/PlumJelly/ABS-Polar

Index Terms— Polarization, polar code, CRC-aided SCL decoder, scaling exponent, BMS channel.

I. INTRODUCTION

POLAR code is the first code family that provably achieves the capacity for all binary-input memoryless symmetric (BMS) channels [1]. In his original paper [1], Arıkan proposed the successive cancellation (SC) decoder and proved that polar codes achieve the capacity of BMS channels under the SC decoder. Later, successive cancellation list (SCL) decoder and CRC-aided SCL decoder [2], [3] were proposed to further reduce the decoding error probability for polar codes. Although polar codes are able to attain the optimal code rate when the code length approaches infinity, the finite-length scaling of polar codes is far from optimal [4], [5], [6], [7]. An extensively-studied method to improve the finite-length performance of polar codes is to replace the \( 2 \times 2 \) Arıkan kernel with large kernels [8], [9], [10], [11], [12], [13]. Recently, a window processing based algorithm was proposed to reduce the decoding complexity of polar codes with large kernels [14], which shows that polar codes with large kernel can be decoded with low decoding complexity. In addition, convolutional polar codes [15] have also been shown to enhance the process of polarization. In a previous paper [16], we proposed the Adjacent-Bits-Swapped (ABS) polar codes which polarize faster than standard polar codes and demonstrate better performance under the CRC-aided SCL decoder. The ABS polar code construction is another way to improve the finite-length performance of polar codes, and its advantages over the large kernel method were discussed in Section II-D of [16]. In this paper, we propose a new family of codes called ABS+ polar codes, which further improve upon ABS polar codes in terms of the polarization speed and the decoding error probabilities.

Standard polar code construction consists of multiple consecutive layers of polar transforms. In the ABS polar code construction, we swap certain pairs of adjacent bits after each layer of polar transform to speed up polarization. Swapping two adjacent bits \( U_i \) and \( U_{i+1} \) can be written as the linear transform \( (U_i, U_{i+1}) \mapsto (U_{i+1}, U_i) \). In this paper, we find that applying the \( 2 \times 2 \) Arıkan transform \( (U_i, U_{i+1}) \mapsto (U_i + U_{i+1}, U_{i+1}) \) to adjacent bits can also accelerate polarization. In total, there are six invertible transforms over the binary field, including the swapping transform \( (U_i, U_{i+1}) \mapsto (U_{i+1}, U_i) \) and the Arıkan transform discussed above. We show that these six invertible transforms are equivalent to (or the same as) the identity transform, the swapping transform, and the Arıkan transform for the purpose of accelerating polarization. Therefore, for each pair of adjacent bits after each layer of polar transform, we only need to decide whether we apply the identity transform, the swapping transform, or the Arıkan transform. In fact, the “+” sign in the name ABS+ polar codes comes from the addition

1In this paper, all additions between binary random variables are over the binary field unless otherwise specified.
in the Arıkan transform. The “+” sign also has an additional meaning that ABS+ polar codes have smaller decoding error probabilities than ABS polar codes.

The encoding matrix $G_{n}^{\text{polar}}$ in standard polar codes is obtained from the recursive relation $G_{n}^{\text{polar}} = G_{n/2}^{\text{polar}} \otimes G_{2}^{\text{polar}}$, where $G_{2}^{\text{polar}} = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$ and $\otimes$ is the Kronecker product. Each Kronecker product is viewed as one layer of polar transform. A standard polar code construction of code length $n = 2^m$ consists of $m$ layers of polar transforms. In the ABS polar code construction, we add a permutation layer after each layer of polar transform, resulting in a different recursive relation $G_{n}^{\text{ABS}} = P_{n}^{\text{ABS}}(G_{n/2}^{\text{ABS}} \otimes G_{2}^{\text{polar}})$. The matrix $P_{n}^{\text{ABS}}$ is an $n \times n$ permutation matrix which swaps certain pairs of adjacent bits to accelerate polarization. For ABS+ polar codes, we use a similar recursive relation $G_{n}^{\text{ABS+}} = Q_{n}^{\text{ABS+}}(G_{n/2}^{\text{ABS+}} \otimes G_{2}^{\text{polar}})$ to construct the encoding matrix, where the $n \times n$ invertible matrix $Q_{n}^{\text{ABS+}}$ performs the swapping transform or the Arıkan transform on certain pairs of adjacent bits.

In the ABS polar code construction, we require that the adjacent bits swapped by the permutation matrix $P_{n}^{\text{ABS}}$ are fully separated. This requirement plays a key role in the efficient decoding of ABS polar codes because it allows us to establish a recursive relation between the joint distribution of every pair of adjacent bits after each layer of polar transform. Similarly, we require that the matrix $Q_{n}^{\text{ABS+}}$ in the ABS+ polar code construction only performs the swapping transform and the Arıkan transform on adjacent bits that are fully separated. As a consequence, a similar recursive relation between the joint distributions of adjacent bits can also be derived for ABS+ polar codes, and the decoding of ABS+ polar codes has the same time complexity as the decoding of ABS polar codes.

We conduct extensive simulations over binary-input AWGN channels to compare the performance of the CRC-aided SCL decoder for ABS+ polar codes, ABS polar codes, and standard polar codes. We run simulations for code length 256, 512, 1024, 2048 and code rates 0.3, 0.5, and 0.7. The decoding time of the following three decoders is more or less the same: (1) standard polar codes with list size 32, (2) ABS polar codes with list size 20, (3) ABS+ polar codes with list size 20. ABS+ polar codes with list size 20 demonstrate 0.1 dB–0.25 dB (respectively, 0.15 dB–0.35 dB) improvement over ABS polar codes with list size 20 (respectively, standard polar codes with list size 32). If we set the list size to be 32 for both ABS+ and standard polar codes, then the decoding time of ABS+ polar codes is 60% longer than that of standard polar codes, but ABS+ polar codes demonstrate 0.2 dB–0.45 dB improvement over standard polar codes.

The rest of this paper is organized as follows: In Section II, we analyze all the possible invertible transform to explain why we only apply the swapping transform or Arıkan transform on a pair of adjacent bits and describe our main idea of the ABS+ polar code construction. In Section III, we give the algorithm to construct ABS+ polar codes. In Section IV, we present the encoding algorithm for ABS+ polar codes and compare ABS+ polar codes with convolutional polar codes. In Section V, we derive the new SC decoder for ABS+ polar codes, and this algorithm can be easily extended to obtain the SCL decoder. Finally, in Section VI, we show our simulation results.

II. MAIN IDEA OF THE ABS+ POLAR CODE CONSTRUCTION

A. The Polarization Framework

Let $U_{1}, U_{2}, \ldots, U_{n}$ be $n$ i.i.d. Bernoulli-1/2 random variables. We view $(U_{1}, \ldots, U_{n})$ as the message vector, and we use an $n \times n$ invertible matrix $G_{n}$ to encode it into the codeword vector $(X_{1}, \ldots, X_{n}) = (U_{1}, \ldots, U_{n})G_{n}$. Each $X_{i}$ is transmitted through a BMS channel $W$, and the channel output vector is denoted as $(Y_{1}, \ldots, Y_{n})$. The SC decoder decodes $U_{i}$ from all the previous message bits $(U_{1}, U_{2}, \ldots, U_{i-1})$ and all the channel output $(Y_{1}, Y_{2}, \ldots, Y_{n})$. For $1 \leq i \leq n$, the conditional entropy

$$H_{i}(G_{n}, W) := H(U_{i}|U_{1}, \ldots, U_{i-1}, Y_{1}, \ldots, Y_{n})$$ (1)

measures the reliability of $U_{i}$ under the SC decoder when we use $G_{n}$ to encode the message vector and transmit the corresponding codeword through the BMS channel $W$. Since $G_{n}$ is invertible, the chain rule of conditional entropy implies that

$$H_{1}(G_{n}, W) + \cdots + H_{n}(G_{n}, W) = n(1 - I(W))$$ (2)

where $I(W)$ is the channel capacity of $W$. A family of matrices $\{G_{n}\}$ is said to be polarizing over a BMS channel $W$ if $H_{i}(G_{n}, W)$ is close to either 0 or 1 for almost all $1 \leq i \leq n$ as $n \to \infty$. It is well-known that if $\{G_{n}\}$ is polarizing, then we can construct a family of capacity-achieving codes from $\{G_{n}\}$. In [16], we quantify the polarization level of an invertible encoding matrix $G_{n}$ over a BMS channel $W$ using the following function

$$\Gamma(G_{n}, W) := \frac{1}{n} \sum_{i=1}^{n} H_{i}(G_{n}, W)(1 - H_{i}(G_{n}, W))$$ (3)

A family of matrices $\{G_{n}\}$ is polarizing over $W$ if $\Gamma(G_{n}, W)$ approaches 0 sufficiently fast as $n$ tends to infinity. Moreover, simulation results indicate that faster convergence rate of $\Gamma(G_{n}, W)$ implies that the corresponding capacity-achieving codes have smaller gap to capacity and better finite-length performance.

The most prominent example of polarizing matrices is $\{G_{n}^{\text{polar}}\}$ in the standard polar code construction, where the polarization level increases after each layer of polar transform, i.e., $\Gamma(G_{n/2}^{\text{polar}} \otimes G_{2}^{\text{polar}}, W) \leq \Gamma(G_{n/2}^{\text{polar}}, W)$. However, each layer of polar transform also increases the code length by a factor of 2. In the ABS polar code construction [16], we find that swapping certain pairs of adjacent bits after each layer of polar transform deepens polarization without increasing the code length. More precisely, we have $\Gamma(P_{n}^{\text{ABS}}(G_{n/2}^{\text{ABS}} \otimes G_{2}^{\text{polar}}), W) \leq \Gamma(G_{n/2}^{\text{ABS}} \otimes G_{2}^{\text{polar}}, W)$, where $P_{n}^{\text{ABS}}$ is a carefully constructed permutation matrix consisting of multiple swaps of disjoint pairs of adjacent bits. In this paper,
we observe that other linear transforms\(^2\) on adjacent bits can also accelerate polarization without increasing code length. There are in total six invertible \(2 \times 2\) linear transforms. In the next subsection, we show that we only need to consider three out of these six transforms, that is, the identity transform, the swapping transform and the Arıkan transform. Choosing one of these three linear transforms for each pair of adjacent bits after each layer of polar transform gives us ABS+ polar codes, which polarize even faster than ABS polar codes.

### B. Classification of Invertible Linear Transforms on Adjacent Bits

As discussed in Section I, we need to construct an invertible matrix \(Q_{\text{ABS}+}\) when we build \(G_{\text{ABS}+}\) from \(G_n^{\text{ABS+}}\) using the recursive relation \(G_{2n}^{\text{ABS+}} = G_{2n/2}^{\text{ABS+}} \otimes G_2^{\text{polar}}\). The matrix \(Q_{\text{ABS}+}\) performs certain linear transforms on certain pairs of adjacent bits. The first step of constructing \(Q_{\text{ABS}+}\) is to find the best linear transform for each fixed pair of adjacent bits. For each \(1 \leq i \leq n-1\), we define \(M_{ij}\) as the set of \(n \times n\) invertible matrices whose corresponding linear transforms map all coordinates except for \(U_i\) and \(U_{i+1}\) in a binary vector \((U_1, U_2, \ldots, U_n)\) to themselves. More precisely, let us write \((U'_1, U'_2, \ldots, U'_n) = (U_1, U_2, \ldots, U_n)M_n\) for an \(n \times n\) invertible matrix \(M_n\) and a binary vector \((U_1, U_2, \ldots, U_n)\). By definition, \(M_n \in M_{ij}\) if and only if \(U'_j = U_j\) for all \(j \in \{1, 2, \ldots, n\} \setminus \{i, i+1\}\) and all \((U_1, U_2, \ldots, U_n) \in \{0, 1\}^n\).

For each \(1 \leq i \leq n-1\), the set \(M_{ij}\) consists of six matrices, denoted as \(I_i, S^{(i)}_n, A^{(i)}_n, D^{(i)}_n, E^{(i)}_n, K^{(i)}_n\). Their corresponding linear transforms are listed below:

1. \(I_i\) corresponds to the identity transform.
2. \(S^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_{i+1}, U_i)\) while fixing all the other \(U_j\)’s unchanged;
3. \(A^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_i + U_{i+1}, U_{i+1})\) while fixing all the other \(U_j\)’s unchanged;
4. \(D^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_i, U_i + U_{i+1})\) while fixing all the other \(U_j\)’s unchanged;
5. \(E^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_{i+1}, U_i + U_{i+1})\) while fixing all the other \(U_j\)’s unchanged;
6. \(K^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_i + U_{i+1}, U_i)\) while fixing all the other \(U_j\)’s unchanged.

Next we show that these six matrices in \(M_{ij}\) can be partitioned into 3 groups of size 2. Two matrices in the same group are equivalent for the purpose of accelerating polarization.

**Lemma 1**: Let \(G_n\) be an \(n \times n\) invertible matrix and let \(W\) be a BMS channel. We have \(\Gamma(D^{(i)}_nG_n, W) = \Gamma(G_n, W)\), \(\Gamma(E^{(i)}_nG_n, W) = \Gamma(S^{(i)}_nG_n, W)\), and \(\Gamma(K^{(i)}_nG_n, W) = \Gamma(A^{(i)}_nG_n, W)\).

**Proof**: Recall the definition of \(H_j(G_n, W)\) in (1). Since the linear transforms corresponding to all six matrices in \(M_{ij}\) map \(U_j\) to itself for every \(j \neq i, i+1\), we have

\[ H_j(G_n, W) = H_j(S^{(i)}_nG_n, W) = H_j(A^{(i)}_nG_n, W) \]

\[ H_j(D^{(i)}_nG_n, W) = H_j(E^{(i)}_nG_n, W) = H_j(K^{(i)}_nG_n, W). \]

Next, we will analyze \(H_i\) and \(H_{i+1}\) in three cases.

1. **(1)** \(D^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_i, U_i + U_{i+1})\). Therefore,
   \[ H_i(D^{(i)}_nG_n, W) = H_i(G_n, W) = H_{i+1}(D^{(i)}_nG_n, W) = H_{i+1}(G_n, W). \]

2. **(2)** \(E^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_{i+1}, U_i + U_{i+1})\), and \(S^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_{i+1}, U_i)\). Therefore,
   \[ H_i(E^{(i)}_nG_n, W) = H_i(G_n, W) = H_{i+1}(E^{(i)}_nG_n, W) = H_{i+1}(G_n, W). \]

3. **(3)** \(K^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_i + U_{i+1}, U_i)\), and \(A^{(i)}_n\) maps \((U_i, U_{i+1})\) to \((U_{i+1}, U_i + U_{i+1})\). Therefore,
   \[ H_i(K^{(i)}_nG_n, W) = H_i(G_n, W) = H_{i+1}(K^{(i)}_nG_n, W) = H_{i+1}(G_n, W). \]

The lemma then follows immediately from definition (3).

Note that \(S^{(i)}_n\) performs the swapping transform on \((U_i, U_{i+1})\), and \(A^{(i)}_n\) performs the Arıkan transform on \((U_i, U_{i+1})\). Lemma 1 tells us that for the purpose of accelerating polarization, we only need to consider the identity transform, the swapping transform, and the Arıkan transform on \((U_i, U_{i+1})\) for each fixed value of \(i\). In fact, there will not be any extra benefit even if we take nonlinear transforms into account because the nonlinear transforms are equivalent to the linear ones in terms of the polarization speed. Indeed, every invertible transform (no matter linear or nonlinear) on two adjacent bits is a permutation on the set \(\{00, 01, 10, 11\}\). Therefore, the total number of invertible transforms on a pair of adjacent bits is \(4! = 24\). For every \(a, b \in \{0, 1\}\), we define an invertible transform \(\{a, b\} = (U_i, U_{i+1}) \mapsto (U_i \oplus a, U_{i+1} \oplus b)\). Then the 24 invertible transforms on \((U_i, U_{i+1})\) are

\[ \bigcup_{a,b \in \{0,1\}} \{ I_i \circ \sigma_i(a,b), S^{(i)}_n \circ \sigma_i(a,b), A^{(i)}_n \circ \sigma_i(a,b), D^{(i)}_n \circ \sigma_i(a,b), E^{(i)}_n \circ \sigma_i(a,b), K^{(i)}_n \circ \sigma_i(a,b) \}. \]
It is clear that the transform \( \sigma_j(a, b) \) does not change \( H_i \) or \( H_{i+1} \) for all \( a, b \in \{0, 1\} \). Therefore, every invertible transform in (4) is equivalent to one of the 3 transforms \( \{I_n, S_n^{(i)}(t), A_n^{(i)}(t)\} \) for the purpose of accelerating polarization. Thus we conclude that using nonlinear transforms on adjacent bits can not further improve the speed of polarization.

Similarly to the ABS polar code construction, we require that \( Q_n^{ABSP} \) only performs the swapping transform, and the Arıkan transform on fully separated pairs of adjacent bits. This fully separated requirement is explained in the next subsection.

C. The Fully Separated Requirement on \( Q_n^{ABSP} \)

We first recall the fully separated requirement on the permutation matrix \( P_n^{ABSP} \) in the ABS polar code construction. Let \( \mathcal{I}^{(n),ABSP} \subseteq \{1, 2, \ldots, n-1\} \) be the set containing the indices of the first bit in each pair of adjacent bits that are swapped by \( P_n^{ABSP} \). Then \( P_n^{ABSP} \) can be written as

\[
P_n^{ABSP} = \prod_{i \in \mathcal{I}^{(n),ABSP}} S_n^{(i)}.
\]

Suppose that \( \mathcal{I}^{(n),ABSP} = \{i_1, i_2, \ldots, i_t\} \), where \( t \) is the size of the set. To enhance the polarization level, we will prove later in Lemma 2 that the elements in \( \mathcal{I}^{(n),ABSP} \) must be even. Furthermore, in order to track the evolution of adjacent-bits-channels, we can not apply transforms on two successive pairs of adjacent bits \( (U_2, U_{2+i}) \) and \( (U_{2+i}, U_{2+i+3}) \) as stated in [16, Section III-G]. The fully separated requirement on \( P_n^{ABSP} \) stipulates that

\[
i_2 \geq i_1 + 4, \quad i_3 \geq i_2 + 4, \quad i_4 \geq i_3 + 4, \quad \ldots, \quad i_t \geq i_{t-1} + 4.
\]

This requirement allows us to track the joint distribution of every pair of adjacent bits through different layers of polar transforms.

A similar requirement is also imposed on \( Q_n^{ABSP} \) in the ABS+ polar code construction. Let \( \mathcal{I}_S^{(n)} \subseteq \{1, 2, \ldots, n-1\} \) be the set containing the indices of the first bit in each pair of adjacent bits upon which \( Q_n^{ABSP} \) performs the swapping transform. Let \( \mathcal{I}_A^{(n)} \subseteq \{1, 2, \ldots, n-1\} \) be the set containing the indices of the first bit in each pair of adjacent bits upon which \( Q_n^{ABSP} \) performs the Arıkan transform. Then \( Q_n^{ABSP} \) can be written as

\[
Q_n^{ABSP} = \left( \prod_{i \in \mathcal{I}_S^{(n)}} S_n^{(i)} \right) \cdot \left( \prod_{i \in \mathcal{I}_A^{(n)}} A_n^{(i)} \right).
\]

Define the set \( \mathcal{I}^{(n)} = \mathcal{I}_S^{(n)} \cup \mathcal{I}_A^{(n)} \), and we write the elements of \( \mathcal{I}^{(n)} \) as \( \mathcal{I}^{(n)} = \{i_1, i_2, \ldots, i_t\} \), where \( t = |\mathcal{I}^{(n)}| \). The fully separated requirement on \( Q_n^{ABSP} \) stipulates that (i) \( \mathcal{I}^{(n)} \cap \mathcal{I}_A^{(n)} = \emptyset \); (ii) the elements in the set \( \mathcal{I}^{(n)} \) satisfy (5). This requirement guarantees that ABS+ polar codes have the same decoding time complexity as ABS polar codes. In Section III-A, we will further prove that all the elements in the set \( \mathcal{I}^{(n)} \) are even numbers.

As a final remark, we need to choose \( m \) matrices \( Q_2^{ABSP}, Q_4^{ABSP}, \ldots, Q_n^{ABSP} \) one by one in the construction of ABS+ polar codes with code length \( n = 2^m \).

III. CODE CONSTRUCTION OF ABS+ POLAR CODES

The first step to construct ABS+ polar codes with code length \( n = 2^m \) is to choose \( m \) matrices \( Q_2^{ABSP}, Q_4^{ABSP}, \ldots, Q_n^{ABSP} \). After that, we calculate the conditional entropies \( H_i(G_i^{ABSP}+W)^n \) and use them to determine which bits are information bits. Since the swapping transform and the Arıkan transform are applied to certain pairs of adjacent bits after each layer of polar transform, there is no recursive relation between bit-channels like the one in standard polar codes. In ABS polar codes [16], we introduced the concept of adjacent-bits-channels to overcome this issue. More precisely, the fully separated requirement (5) on \( P_n^{ABSP} \) allows us to establish a recursive relation between adjacent-bits-channels. We organize this section as follows: In Section III-A, we introduce the ADB polar transform and establish the recursive relation for ABS+ polar codes. In Section III-B, we describe how to choose the matrices \( Q_2^{ABSP}, Q_4^{ABSP}, \ldots, Q_n^{ABSP} \). Finally, in Section III-C, we summarize the algorithm of the ABS+ polar code construction.

A. Recursive Relation Between Adjacent-Bits-Channels

In this subsection, we describe how to calculate the conditional entropies \( H_i(G_i^{ABSP}+W)^n \) for ABS+ polar codes in Fig. 1. Since \( H_i(G_i^{ABSP}+W)^n = 1 - I(W_i^{(n),ABSP}) \), we only need to calculate the transition probabilities of \( W_i^{(n),ABSP} \) for ABS+ polar codes.

In standard polar codes, the transition probabilities of bit-channels are calculated using a recursive relation which is not available for ABS or ABS+ polar codes. Following the method in [16], we define the adjacent-bits-channels

\[
\{W_i^{(n),ABSP} \}_{i=1}^{n-1} \quad \text{for ABS+ polar codes in Fig. 1, and we will derive a recursive relation between } \{W_i^{(n),ABSP} \}_{i=1}^{n-1} \text{ and } \{y_i^{(n,2),ABSP} \}_{i=1}^{n/2-1}.
\]

Once we obtain the transition probabilities of \( \{W_i^{(n),ABSP} \}_{i=1}^{n-1} \) from this recursive relation, the transition probabilities of \( \{W_i^{(n),ABSP} \}_{i=1}^{n-1} \) can be calculated as follows:

\[
W_i^{(n),ABSP}(y_1, \ldots, y_n, u_1, \ldots, u_{i-1}, u_i) = \frac{1}{2} \sum_{u_{i+1} \in \{0,1\}} W_{i+1}^{(n),ABSP}(y_1, \ldots, y_n, u_1, \ldots, u_{i-1}, u_i, u_{i+1}),
\]

\[
W_{i+1}^{(n),ABSP}(y_1, \ldots, y_n, u_1, \ldots, u_{i}, u_{i+1}) = \frac{1}{2} W_i^{(n),ABSP}(y_1, \ldots, y_n, u_1, \ldots, u_{i-1}, u_i, u_{i+1}).
\]
for $1 \leq i \leq n-1$. In Fig. 1, we also introduce bit-channels $\{\hat{W}_i^{(n),\text{ABS}+}\}_{i=1}^n$ and adjacent-bits-channels $\{\hat{V}_i^{(n),\text{ABS}+}\}_{i=1}^n$, which represent the channels prior to applying a layer of invertible transforms $Q_n^{\text{ABS}+}$ to specific pairs of adjacent bits.

As mentioned at the end of Section II-C, the set $T^{(n),\text{ABS}}$ in ABS polar codes and $T^{(n)}$ in ABS+ polar codes both satisfy the requirement (5), which guarantees the existence of a recursive relation between $\{V_i^{(n),\text{ABS}+}\}_{i=1}^{n-1}$ and $\{V_i^{(n),\text{ABS}+}\}_{i=1}^{n/2-1}$. The necessity of the requirement (5) was explained in [16, Section III-G]. Here we point out another similarity between the two sets $T^{(n),\text{ABS}}$ and $T^{(n)}$. In [16, Section III-C], we showed that all the elements in $T^{(n),\text{ABS}}$ are even numbers. Next we prove that all the elements in $T^{(n)}$ are also even numbers. Following the notation in Fig. 1, we only need to show that applying the swapping transform or the Arıkan transform to $(\hat{U}_{2j-1}, \hat{U}_{2j})$ does not increase the polarization level for any $1 \leq j \leq n/2$. This is proved in Lemma 2 below.

Recall the definitions of $S_n^{(i)}$ and $A_n^{(i)}$ in Section II-B. 

**Lemma 2:** Let $W$ be a BMS channel. We use the shorthand notation $G_n^{\text{ABS}+} = G_{n/2}^{\text{ABS}+} \otimes G_2^{\text{polar}}$. For $1 \leq j \leq n/2$, we have

$$\Gamma(S_n^{(2j-1),G_n^{\text{ABS}+}}, W) = \Gamma(A_n^{(2j-1),G_n^{\text{ABS}+}}, W) \geq \Gamma(\hat{W}^{\text{ABS}+}, W).$$

(8)

**Proof:** Recall the definition of $H_i(G_n, W)$ in (1). Clearly, for every $i \neq 2j-1, 2j$, we have

$$H_i(S_n^{(2j-1),G_n^{\text{ABS}+}}, W) = H_i(A_n^{(2j-1),G_n^{\text{ABS}+}}, W) = H_i(\hat{W}^{\text{ABS}+}, W).$$

(9)

Next we analyze $H_{2j-1}$ and $H_{2j}$. Following the notation in Fig. 1,

$$H_{2j-1}(\hat{W}^{\text{ABS}+}, W) = H(\hat{U}_{2j}, \hat{U}_{2j-2}, Y_1, Y_2, \ldots, Y_n)$$

(10)

$$H_{2j}(\hat{W}^{\text{ABS}+}, W) = H(\hat{U}_{2j}, \hat{U}_{2j-1}, Y_1, Y_2, \ldots, Y_n).$$

By definition,

$$H_{2j-1}(S_n^{(2j-1),\hat{W}^{\text{ABS}+}}, W) = H(\hat{U}_{2j}, \hat{Y}_{2j-1}, \hat{Y}_{2j})$$

$$= H(\hat{X}_{2j}, \hat{Y}_{2j-1}, \hat{Y}_{2j}) = H(\hat{X}_{2j}) = 1 - I(W_j^{(n/2),\text{ABS}+}).$$

(11)
Then we have
\[ H_2_j(S_n^{(2j-1)}G_n^{ABS+}, W) = H(\tilde{U}_2_j-1|\tilde{Y}_2_j-1, \tilde{Y}_2_j) \]
\[ = H(\tilde{X}_2_j-1 + \tilde{X}_2_j|\tilde{Y}_2_j-1, \tilde{Y}_2_j) = H(\tilde{X}_2_j-1|\tilde{X}_2_j, \tilde{Y}_2_j-1, \tilde{Y}_2_j) \]
\[ = H(\tilde{X}_2_j-1|\tilde{Y}_2_j-1) = 1 - I(W_j^{(n/2)}, ABS^+). \]

Similarly,
\[ H_{2j-1}(A_n^{(2j-1)}G_n^{ABS+}, W) = H(\tilde{U}_2_j-1 + \tilde{U}_2_j|\tilde{Y}_2_j-1, \tilde{Y}_2_j) \]
\[ = H(\tilde{X}_2_j-1|\tilde{Y}_2_j-1) = H(\tilde{X}_2_j-1|\tilde{X}_2_j, \tilde{Y}_2_j-1, \tilde{Y}_2_j) \]
\[ = H(\tilde{X}_2_j|\tilde{X}_2_j-1, \tilde{Y}_2_j-1, \tilde{Y}_2_j) \]
\[ = H(\tilde{X}_2_j|\tilde{Y}_2_j) = 1 - I(W_j^{(n/2)}, ABS^+). \]

Then we have
\[ H_{2j-1}(S_n^{(2j-1)}G_n^{ABS+}, W) = H_{2j}(S_n^{(2j-1)}G_n^{ABS+}, W) \]
\[ = H_{2j-1}(A_n^{(2j-1)}G_n^{ABS+}, W) = H_{2j}(A_n^{(2j-1)}G_n^{ABS+}, W) \]
\[ = 1 - I(W_j^{(n/2)}, ABS^+). \]

Combining this with (9) and the definition of \( \Gamma(G_n, W) \) in (3), we prove the first equality in (8).

According to (11), we have
\[ H_{2j-1}(\tilde{G}_{n}^{ABS^+}, W)(1 - H_{2j-1}(\tilde{G}_{n}^{ABS^+}, W)) + H_{2j}(\tilde{G}_{n}^{ABS^+}, W)(1 - H_{2j}(\tilde{G}_{n}^{ABS^+}, W)) \]
\[ = H_{2j-1}(\tilde{G}_{n}^{ABS^+}, W) + H_{2j}(\tilde{G}_{n}^{ABS^+}, W) \]
\[ - (H_{2j-1}(\tilde{G}_{n}^{ABS^+}, W))^2 - (H_{2j}(\tilde{G}_{n}^{ABS^+}, W))^2 \]
\[ = 2(1 - I(W_j^{(n/2)}, ABS^+)) - (H_{2j-1}(\tilde{G}_{n}^{ABS^+}, W))^2 \]
\[ - (H_{2j}(\tilde{G}_{n}^{ABS^+}, W))^2 \]
\[ \leq 2(1 - I(W_j^{(n/2)}, ABS^)) - \frac{1}{2} (H_{2j-1}(\tilde{G}_{n}^{ABS^+}, W)) \]
\[ + H_{2j}(\tilde{G}_{n}^{ABS^+}, W) \]
\[ = 2(1 - I(W_j^{(n/2)}, ABS^)) - 2(1 - I(W_j^{(n/2)}, ABS^))^2 \]
\[ = 2(1 - I(W_j^{(n/2)}, ABS^))I(W_j^{(n/2)}, ABS^) \]
\[ = H_{2j-1}(A_n^{(2j-1)}G_n^{ABS+}, W)(1 - H_{2j-1}(A_n^{(2j-1)}G_n^{ABS+}, W)) + H_{2j}(A_n^{(2j-1)}G_n^{ABS+}, W)(1 - H_{2j}(A_n^{(2j-1)}G_n^{ABS+}, W)), \]

(13)

where the inequality follows from the Cauchy–Schwarz inequality, and the last equality follows from (12). Combining (13) with (9), we prove the inequality in (8).

The DB polar transform and the SDB polar transform (see Fig. 3 for their definitions) were introduced in [16]. Section III] to describe the recursive relation for ABS polar codes. In this paper, we introduce a new transform called the Added-Double-Bits (ADB) polar transform to characterize the recursive relation for ABS polar codes. The DB polar transform corresponds to applying the identity transform on a pair of adjacent bits, the SDB polar transform corresponds to applying the swapping transform on a pair of adjacent bits, and the ADB polar transform corresponds to applying the Arikan transform on a pair of adjacent bits.

The details of the ADB polar transform are illustrated in Fig. 4.

Given an adjacent-bits-channel \( V \), the transition probabilities of \( V^\triangledown : \{0, 1\}^2 \rightarrow \{0, 1\} \), \( V^\cdot : \{0, 1\}^2 \rightarrow \{0, 1\}^2 \), and \( V^\Delta : \{0, 1\}^2 \rightarrow \{0, 1\}^2 \times \{0, 1\} \) in Fig. 4 are given by

\[ V^\triangledown(y_1, y_2|u_1, u_2) = \frac{1}{4} \sum_{u_3, u_4 \in \{0, 1\}} V(y_1|u_1 + u_2 + u_3), \]
\[ u_3 + u_4 V(y_2|u_2 + u_3, u_4) \]
for \( u_1, u_2 \in \{0, 1\} \) and \( y_1, y_2 \in \mathcal{Y} \),
\[ V^\circ \left( u_1, y_1, y_2 | u_2, u_3 \right) = \frac{1}{4} \sum_{u_4 \in \{0, 1\}} V(y_1|u_1 + u_2 + u_3, \]
\[ u_3 + u_4 V(y_2|u_2 + u_3, u_4) \]
for \( u_1, u_2, u_3 \in \{0, 1\} \) and \( y_1, y_2 \in \mathcal{Y} \),
\[ V^\triangle \left( u_1, u_2, y_1, y_2 | u_3 \right) = \frac{1}{4} \left( V(y_1|u_1 + u_2 + u_3, \]
\[ u_3 + u_4 V(y_2|u_2 + u_3, u_4) \]
for \( u_1, u_2, u_3 \in \{0, 1\} \) and \( y_1, y_2 \in \mathcal{Y} \). (15)

Now we are ready to state the recursive relation between \( \{V_i^{(n)}, ABS+\}_{i=1}^{n-1} \) and \( \{V_i^{(n/2)}, ABS+\}_{i=1}^{n/2-1} \).

**Lemma 3**: Let \( n \geq 4 \). We write \( Q_{n/2}^{ABS+} \) in the form of (6) and require that \( \mathcal{T}^{(n)} = \{2j_1, 2j_2, \ldots, 2j_t\} \) satisfies (14). For \( 1 \leq j \leq n/2 - 1 \), we have the following results:

**Case i)** If \( 2j \in \mathcal{T}_S^{(n)} \), then
\[ \begin{align*}
V_{2j-1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\downarrow, \\
V_{2j}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\circ, \\
V_{2j+1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\triangle.
\end{align*} \]

**Case ii)** If \( 2j \in \mathcal{T}_A^{(n)} \), then
\[ \begin{align*}
V_{2j-1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\triangledown, \\
V_{2j}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\circ, \\
V_{2j+1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\triangle.
\end{align*} \]

**Case iii)** If \( 2(j-1) \in \mathcal{T}_S^{(n)} \), \( 2(j+1) \notin \mathcal{T}^{(n)} \), then
\[ V_{2j}^{(n), ABS+} = (V_{j}^{(n/2), ABS+})^\circ. \]

**Case iv)** If \( 2(j-1) \notin \mathcal{T}_S^{(n)} \), \( 2(j+1) \notin \mathcal{T}^{(n)} \), then
\[ \begin{align*}
V_{2j-1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\triangledown, \\
V_{2j}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\circ. \\
V_{2j+1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\triangle.
\end{align*} \]

**Case v)** If \( 2(j-1) \notin \mathcal{T}_S^{(n)} \), \( 2j \notin \mathcal{T}^{(n)} \), then
\[ V_{2j-1}^{(n), ABS+} = (V_{j}^{(n/2), ABS+})^\triangledown, \]
\[ V_{2j}^{(n), ABS+} = (V_{j}^{(n/2), ABS+})^\circ. \]

**Case vi)** If \( 2(j-1) \notin \mathcal{T}_S^{(n)} \), \( 2j \notin \mathcal{T}^{(n)} \), then
\[ \begin{align*}
V_{2j-1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\triangledown, \\
V_{2j}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\circ, \\
V_{2j+1}^{(n), ABS+} &= (V_{j}^{(n/2), ABS+})^\triangle.
\end{align*} \]

We omit the proof of Lemma 3 because it is similar to the proof of [16, Lemma 1, Lemma 2].

For a given BMS channel \( W \), the starting point of the recursive relation in Lemma 3 is \( V_1^{(2), ABS+} \), whose transition probabilities can be calculated as follows
\[ V_1^{(2), ABS+}(y_1, y_2 | u_1, u_2) = W(y_1|u_1 + u_2)W(y_2|u_2) \]
for \( u_1, u_2 \in \{0, 1\} \) and \( y_1, y_2 \in \mathcal{Y} \).

(16)

After calculating the transition probabilities of the adjacent-bits-channels \( \{V_i^{(n), ABS+}\}_{i=1}^{n-1} \), we can use (7) to obtain the transition probabilities of the bit-channels \( \{W_i^{(n), ABS+}\}_{i=1}^{n} \). This allows us to calculate \( H_1(G_{n/2}^{ABS+}, W) \) and determine which bits are information bits.

**B. Constructing the Matrices \( Q_2^{ABS+}, Q_2^{ABS+}, \ldots, Q_n^{ABS+} \)**

We construct \( Q_{2k}^{ABS+}, Q_{2k}^{ABS+}, \ldots, Q_{2k}^{ABS+} \) one by one, starting from \( Q_2^{ABS+} \). Therefore, the matrices \( Q_{2k}^{ABS+}, Q_{2k}^{ABS+}, \ldots, Q_{2k}^{ABS+} \) and \( Q_{2k}^{ABS+} \) are already known when we construct \( Q_{2k+1}^{ABS+} \). Lemma 3 allows us to calculate the transition probabilities of the adjacent-bits-channels \( \{V_i^{(n/2), ABS+}\}_{i=1}^{n/2-1} \) from \( Q_{2k}^{ABS+}, Q_{2k}^{ABS+}, \ldots, Q_{2k}^{ABS+} \), so we also know the transition probabilities of \( \{V_i^{(n/2), ABS+}\}_{i=1}^{n/2-1} \) when constructing \( Q_{2k+1}^{ABS+} \). Constructing the matrix \( Q_k^{ABS+} \) is equivalent to constructing the two sets \( T_S^{(n)} \) and \( T_A^{(n)} \) in (6), whose elements are all even numbers.

According to the recursive relation \( G_n^{ABS+} = G_n^{ABS+}(G_{n/2}^{ABS+} \otimes G_{2}^{polar}) \), our objective is to choose \( T_S^{(n)} \) and \( T_A^{(n)} \) whose corresponding matrix \( Q^{ABS+} \) minimizes \( H(G_n^{ABS+}, W) \) for a given \( G_n^{ABS+} \). This is equivalent to maximizing \( H(G_{n/2}^{ABS+} \otimes G_2^{polar}, W) \) − \( H(G_n^{ABS+}, W) \). To that end, let us introduce some notation. Suppose that \( V : \{0, 1\}^2 \rightarrow \mathcal{Y} \) is an adjacent-bits-channel. Let two i.i.d. Bernoulli-1/2 random variables \( U_1 \) and \( U_2 \) be the inputs of \( V \), and let \( Y \) be the corresponding channel output. Define
\[ \gamma(V) = H(U_1|Y) - (1 - H(U_1|Y)) \]
\[ + H(U_2|U_1, Y)(1 - H(U_2|U_1, Y)). \]

Recall the definitions of \( \{V_i^{(n), ABS+}\}_{i=1}^{n-1} \) and \( \{V_i^{(n), ABS+}\}_{i=1}^{n-1} \) in Fig. 1. It is easy to see that
\[ \Gamma(G_n^{ABS+}, W) = \frac{1}{n} \left[H_1(G_n^{ABS+}, W)(1 - H_1(G_n^{ABS+}, W)) \right. \]
\[ + \sum_{j=1}^{n/2-1} \gamma(V_j^{(n), ABS+}) \]
\[ \left. + H_n(G_n^{ABS+}, W)(1 - H_n(G_n^{ABS+}, W)) \right]. \]

(17)
Lemma 3 implies that
\[
\hat{\Gamma}^{(n)}(n).\text{ABS} = \left\{ \begin{array}{ll}
(V_j^{(n/2)},{\text{ABS}}^+) & \text{if } 2j \in \mathcal{I}^{(n)}_S, \\
(V_j^{(n/2)},{\text{ABS}}^+) & \text{if } 2j \in \mathcal{I}^{(n)}_A, \\
(V_j^{(n/2)},{\text{ABS}}^+) & \text{otherwise}.
\end{array} \right.
\]
Taking this into (17), we obtain that
\[
\Gamma(G_{n/2}^{\text{ABS}} \otimes G_2^{\text{polar}}, W) - \Gamma(G_n^{\text{ABS}}, W)
\]
\[
= \frac{1}{n} \sum_{2j \in \mathcal{I}^{(n)}_S} \left[ \gamma((V_j^{(n/2)},{\text{ABS}}^+)\hat{\circ} - \gamma((V_j^{(n/2)},{\text{ABS}}^+)) \right]
\]
\[
\quad + \sum_{2j \in \mathcal{I}^{(n)}_A} \left[ \gamma((V_j^{(n/2)},{\text{ABS}}^+)) - \gamma((V_j^{(n/2)},{\text{ABS}}^+)) \right]
\]
\[
= \frac{1}{n} \sum_{2j \in \mathcal{I}^{(n)}} \text{score}(j).
\]
Therefore, to construct \( \mathcal{I}^{(n)}_S \) and \( \mathcal{I}^{(n)}_A \), we only need to find their union \( \mathcal{I}^{(n)} \). Next we define the function
\[
\text{score}(j) = \max \left\{ \gamma((V_j^{(n/2)},{\text{ABS}}^+)\hat{\circ} - \gamma((V_j^{(n/2)},{\text{ABS}}^+)), \right.
\]
\[
\left. \gamma((V_j^{(n/2)},{\text{ABS}}^+)) - \gamma((V_j^{(n/2)},{\text{ABS}}^+)) \right\}
\]
for \( 1 \leq j \leq n/2 - 1 \). Taking (19), (20) into (18), we obtain that
\[
\Gamma(G_{n/2}^{\text{ABS}} \otimes G_2^{\text{polar}}, W) - \Gamma(G_n^{\text{ABS}}, W)
\]
\[
= \frac{1}{n} \sum_{2j \in \mathcal{I}^{(n)}} \text{score}(j).
\]
Therefore, we need to find \( \mathcal{I}^{(n)} \) to maximize the right-hand side of (21) under the constraint (14), i.e., we need to solve the following optimization problem:
\[
\mathcal{I}^{(n)} = \arg\max_{S \subseteq \{2,4,...,n-2\}} \sum_{2j \in S} \text{score}(j)
\]
\[
\text{s.t. } |j_1 - j_2| \geq 2 \text{ for all distinct } j_1, j_2 \in S. \tag{22}
\]
This problem can be solved using a dynamic programming method with time complexity \( O(n) \). More precisely, for \( k \in \{2,4,...,n-2\} \), we define
\[
\mathcal{I}^{(n)}_k = \arg\max_{S \subseteq \{2,4,...,k\}} \sum_{2j \in S} \text{score}(j)
\]
\[
\text{s.t. } |j_1 - j_2| \geq 2 \text{ for all distinct } j_1, j_2 \in S. \tag{23}
\]
Note that \( \mathcal{I}^{(n)} = \mathcal{I}^{(n-2)} \). The sets \( \mathcal{I}^{(n)}_2, \mathcal{I}^{(n)}_4, \ldots, \mathcal{I}^{(n)}_{n-2} \) can be calculated from the following recursive relation
\[
\mathcal{I}^{(n)}_k = \left\{ \begin{array}{ll}
\mathcal{I}^{(n-2)}_k & \text{if } k \leq n/2 - 1 \\
\mathcal{I}^{(n-2)}_k & \text{otherwise}
\end{array} \right.
\]
\[
\mathcal{I}^{(n)}_k = \left\{ \begin{array}{ll}
\mathcal{I}^{(n-2)}_k \cup \{k+2\} & \text{if } \sum_{2j \in \mathcal{I}^{(n-2)}_k } \text{score}(j) > \sum_{2j \in \mathcal{I}^{(n-2)}_k } \text{score}(j) \text{ otherwise}
\end{array} \right.
\]
\[
\mathcal{I}^{(n)}_1 = \emptyset, \quad \mathcal{I}^{(n)}_2 = \{2\} \text{ if } \text{score}(1) > 0 \tag{24}
\]
In this way, we solve the optimization problem (22). Finally, we use (19) to obtain \( \mathcal{I}^{(n)}_S \) and \( \mathcal{I}^{(n)}_A \).

C. Summary of the ABS+ Polar Code Construction

In the previous subsections, we describe two main ingredients of the ABS+ polar code construction. The first ingredient is the method to recursively calculate the transition probabilities of \( \{V_i^{(n)},{\text{ABS}}^+\}_{i=1}^{n-1} \) when \( Q_2^{\text{ABS}^+}, Q_4^{\text{ABS}^+}, \ldots, Q_n^{\text{ABS}^+} \) are known. The second ingredient is the algorithm to construct the matrix \( Q_n^{\text{ABS}^+} \) when the transition probabilities of \( \{V_i^{(n)},{\text{ABS}}^+\}_{i=1}^{n-1} \) are available. Moreover, we also need to quantize the output alphabets using Algorithm 1 in [16] to ensure that the output alphabet size of \( \{V_i^{(n)},{\text{ABS}}^+\}_{i=1}^{n-1} \) does not increase exponentially in \( n \). Below we put everything together and summarize the ABS+ polar code construction in Algorithm 1.

IV. THE ENCODING ALGORITHM FOR ABS+ POLAR CODES

In this section, we describe the encoding algorithm of ABS+ polar codes and give an example of an ABS+ polar code with code length \( n = 16 \). We will also use this example to illustrate how our new SC decoder works in Section V.

Let \( C \) be an \( (n,k) \) ABS+ polar code defined by the matrices \( Q_2^{\text{ABS}^+}, Q_4^{\text{ABS}^+}, \ldots, Q_n^{\text{ABS}^+} \) (or equivalently, defined by the sets \( \mathcal{I}^{(2)}_S, \mathcal{I}^{(2)}_A, \mathcal{I}^{(4)}_S, \mathcal{I}^{(4)}_A, \ldots, \mathcal{I}^{(n)}_S, \mathcal{I}^{(n)}_A \)). Let \( A = \{1,2,\ldots,i,k\} \) be the index set of the information bits in \( C \). We present the encoding algorithm of the code \( C \) in Algorithm 2.

Proposition 1: The time complexity of Algorithm 2 is \( O(n \log(n)) \).

Apart from Lines 7-11, the rest of Algorithm 2 is the same as the encoding algorithm of standard polar codes, whose time complexity is \( O(n \log(n)) \). According to (6), the operations in Lines 7-11 are equivalent to
\[
(c_h, c_{h+1}, c_{h+2}, \ldots, c_{h+(n_c-1)}) = (c_h, c_{h+1}, c_{h+2}, \ldots, c_{h+(n_c-1)})Q_{n_c}^{\text{ABS}^+}.
\]
The fully separated requirement (14) in our code construction guarantees that each element in the vector \( \{c_h, c_{h+1}, c_{h+2}, \ldots, c_{h+(n_c-1)}\} \) is involved in at most one swapping transform or one Arıkan transform. Therefore, the number of operations in Lines 7-11 is no more than \( n_c = \binom{n}{2} - 1 \). For each fixed value of \( i \), Lines 7-11 are executed \( t = 2^i \) times and induce at most \( n_c \cdot t = n \) operations. Since the value
Thus we conclude that the time complexity of Algorithm 2 is of operations in Lines 7-11 is upper bounded by \( n \) of (16).

We apply the swapping transform on the pairs \((U_8, U_9)\) because \( I_S^{(16)} = \{8, 12\} \). We apply the Arıkan transform on the pairs \((X_2^{(4)}, X_9^{(4)})\), \((X_6^{(4)}, X_{10}^{(4)})\), \((X_7^{(4)}, X_{11}^{(4)})\), and \((X_8^{(4)}, X_{12}^{(4)})\) because \( I_A^{(8)} = \{4\} \). We apply the swapping transform on the pairs \((X_2^{(4)}, X_9^{(4)})\), \((X_6^{(4)}, X_{10}^{(4)})\), \((X_7^{(4)}, X_{11}^{(4)})\), and \((X_8^{(4)}, X_{12}^{(4)})\) because \( I_S^{(4)} = \{2\} \).

of \( i \) ranges from 0 to \( \log(n) - 1 \) in Line 3, the total number of operations in Lines 7-11 is upper bounded by \( n \log(n) \). Thus we conclude that the time complexity of Algorithm 2 is \( O(n \log(n)) \).

Next we give a concrete example of an \((n = 16, k = 8)\) ABS+ polar code defined by the following sets:

\[
\begin{align*}
I_S^{(2)} &= \emptyset, \\
I_S^{(4)} &= \{2\}, \\
I_S^{(8)} &= \emptyset, \\
I_S^{(16)} &= \{4\}, \\
I_A^{(2)} &= \emptyset, \\
I_A^{(4)} &= \emptyset, \\
I_A^{(8)} &= \{4\}, \\
I_A^{(16)} &= \{8, 12\}, \\
A &= \{9, 10, 11, 12, 13, 14, 15, 16\}.
\end{align*}
\]

(25)
The encoding circuit of this specific ABS+ polar code is given in Fig. 5.

We can see from Fig. 5 that the structure of ABS+ polar codes bears some resemblance to that of convolutional polar codes. Both approaches deepen the polarization level by applying invertible transforms to adjacent bits. However, they differ in two aspects. Firstly, ABS+ polar codes consider all invertible transforms that can enhance the polarization level, whereas convolutional polar codes only employ the Arıkan transform on adjacent bits. Additionally, ABS+ polar codes strictly limit the number of adjacent bit pairs participating in the invertible transforms, while convolutional polar codes apply the Arıkan transform to all adjacent bits regardless of whether it enhances the polarization level of the current pair of adjacent bits or not. In terms of implementation, convolutional polar codes require tracking the joint distribution of successive three bits at each layer in order to apply the Arıkan transformation to all adjacent bits. For ABS+ polar codes, if we also track the joint distribution of each successive three bits, we can ignore the fully separated requirement (5) and apply invertible transforms to more adjacent bits. This can further enhance the polarization level but also increase the algorithm’s complexity. If we refer to the ABS+ polar codes with ignored fully separated requirement (5) as extended ABS+ polar codes, then convolutional polar codes can be considered as a special case of extended ABS+ polar codes, as latter exhibit higher flexibility in determining when and which transform to apply.

V. THE SC DECODING ALGORITHM FOR ABS+ POLAR CODES

Although the SCL decoder is more widely used in practice, we will only describe the SC decoder in this paper for the sake of simplicity. The method of extending the SC decoder to obtain the SCL decoder is well-known in the polar coding literature: see [2] and [16] for example.

We will first present a version of the SC decoder with space complexity \( O(n \log(n)) \). This version is relatively easy to understand. Then in Section V-A, we present a space-efficient version with space complexity \( O(n) \).

Recall that \((U_1, U_2, \ldots, U_n)\) is the message vector, and \((X_1, X_2, \ldots, X_n)\) is the codeword vector. Following the example in Fig. 5, we define some intermediate vectors \(\{(X_1^{(m)}, X_2^{(m)}, \ldots, X_n^{(m)})\}_{n=2, 4, \ldots, n}\). Let \((X_1^{(n)}, X_2^{(n)}, \ldots, X_n^{(n)}) = (U_1, U_2, \ldots, U_n)\). The intermediate vectors are defined recursively...
from $n_c = n/2$ to $n_c = 2$ using the following relation

\[ (X_1^{(n_c)}, X_2^{(n_c)}, \ldots, X_{n/2}^{(n_c)}) = (Q_{2n_1}^{n_1} (I_{n_c} \otimes G_2^{\text{Polar}})) \otimes I_{n/(2n_1)}. \]  

It is easy to see that the $n/2$ random vectors

\[ \left\{ (X_1^{(n_c)}, X_2^{(n_c)}, \ldots, X_{n/2}^{(n_c)}) \right\}_{\beta=1}^{n/2}, \]

are independent and identically distributed. For $1 \leq i \leq n_c - 1$ and $1 \leq \beta \leq n/n_c$, we define two random vectors

\[ X_i^{(n_c)} = (X_1^{(n_c)}, X_2^{(n_c)}, \ldots, X_{\beta-1}^{(n_c)}, X_\beta^{(n_c)}), \]

\[ Y_i^{(n_c)} = (Y_1^{(n_c)}, Y_2^{(n_c)}, \ldots, Y_{\beta-1}^{(n_c)}). \]

According to the definition in Fig. 1, the channel mapping from $(X_1^{(n_c)}, X_2^{(n_c)}, \ldots, X_{\beta-1}^{(n_c)}, X_\beta^{(n_c)}$ to $(X_1^{(n_c)}, Y_\beta^{(n_c)})$ is the adjacent-bits-channel $V_i^{(n_c), \text{ABS}^+}$ for all $1 \leq \beta \leq n/n_c$.

Below we omit “ABS+” in the superscript and simply write $V_i^{(n_c)}$.

Let $(y_1, \ldots, y_n)$ be a realization of the channel output random vector, i.e., $(y_1, \ldots, y_n)$ is the input to the SC decoder. For $n_c = 2, 4, \ldots, n$, let $(\hat{x}_1^{(n_c)}, \ldots, \hat{x}_n^{(n_c)})$ be the decoding result of $(X_1^{(n_c)}, \ldots, X_n^{(n_c)})$ given by the SC decoder. Similarly, let $(\hat{u}_1, \ldots, \hat{u}_n)$ be the decoding result of $(U_1, \ldots, U_n)$.

The SC decoder determines the value of $U_i$ from the conditional probabilities

\[ \mathbb{P}(U_i = \hat{u}_1, \ldots, U_{i-1} = \hat{u}_{i-1}, Y_1 = y_1, \ldots, Y_n = y_n | U_i = u_i, U_{i+1} = u_{i+1}) \]

\[ = \mathbb{P}(u_i, u_{i+1} \in \{0, 1\}). \]  

For $1 \leq i \leq n_c - 1$ and $1 \leq \beta \leq n/n_c$, we write

\[ \hat{x}_i^{(n_c)} = (\hat{x}_1^{(n_c)}, \hat{x}_2^{(n_c)}, \ldots, \hat{x}_\beta^{(n_c)}), \]

\[ y_\beta^{(n_c)} = (y_\beta, y_{\beta+2n/n_c}, \ldots, y_{\beta+(n-1)n/n_c}). \]

The SC decoder calculates the conditional probabilities in (28) recursively from

\[ \mathbb{P}(X_i^{(n_c)}, Y_\beta^{(n_c)} = (x_i^{(n_c)}, y_\beta^{(n_c)}), | X_{\beta-1}^{(n_c)}(n_c/2 - 1) = a, X_{\beta+(i-1)n/n_c}^{(n_c)} = b) \]

\[ = \mathbb{P}(x_i^{(n_c)}, y_\beta^{(n_c)} | a, b), \quad a, b \in \{0, 1\}. \]
Algorithm 3 ABS+Decode((y₁, y₂, ..., yₙ))

Input: the received vector (y₁, y₂, ..., yₙ) ∈ ℝⁿ
Output: the decoded codeword
(\hat{x}_1, \hat{x}_2, ..., \hat{x}_n) ∈ {0, 1}ⁿ

for β ∈ {1, 2, ..., n/2}, a ∈ {0, 1} and b ∈ {0, 1} do
  \[ P_2[1, \beta][a, b] = W(y_β)a + b \cdot W(y_{\beta+n/2}[b]) \]
enddo

decode_channel(2, 1)  \> Recursive decoding
for β = 1, 2, ..., n do
  \[ \hat{x}_β → B_2[1, \beta] + B_2[2, \beta] \]
endfor
\[ (\hat{x}_1, \hat{x}_2, ..., \hat{x}_n) \]

For each nₑ ∈ {2, 4, 8, ..., n}, we use a data structure Pₙₑ to store the probabilities in (30). More specifically, Pₙₑ is a four-dimensional array with indices i ∈ {1, 2, ..., nₑ - 1}, β ∈ {1, 2, ..., nₑ/nₑ}, a ∈ {0, 1}, b ∈ {0, 1}. We write an entry in Pₙₑ as Pₙₑ[i, β, a, b], which stores V_i^{(nₑ)}(\hat{x}_β, \hat{y}_β) | a, b), i.e.,

\[ Pₙₑ[i, β, a, b] = V_i^{(nₑ)}(\hat{x}_β, \hat{y}_β) | a, b). \quad (31) \]

We omit x_j^{(nₑ)} and y_j^{(nₑ)} in the notation Pₙₑ[i, β, a, b] because they remain unchanged in the whole decoding procedure.

We use another data structure Bₙₑ to store the decoding results of the intermediate vector (X_1^{(nₑ)}, X_2^{(nₑ)}, ..., X_n^{(nₑ)}). The data structure Bₙₑ is a two-dimensional array with indices i ∈ {1, 2, ..., nₑ}, β ∈ {1, 2, ..., nₑ/nₑ}. We write an entry in Bₙₑ as Bₙₑ[i, β], which stores \( \hat{x}_{β+i-1}/nₑ \), i.e.,

\[ Bₙₑ[i, β] = \hat{x}_{β+i-1}/nₑ \quad \beta \in \{1, 2, ..., nₑ \}. \quad (32) \]

Pₙₑ and Bₙₑ are the only two data structures we need in the SC decoder. The number of entries in Pₙₑ is 4(nₑ - nₑ/2), and the number of entries in Bₙₑ is nₑ. Since nₑ takes log(nₑ) values, the space complexity of the SC decoder is O(nₑ log(nₑ)). In Section V-A, we will show how to reduce the space complexity to O(nₑ).

Algorithm 3 outlines three main steps of the SC decoder. As an initialization, we calculate all the entries in the array P₂ in Lines 1-2, where the formula in Line 2 follows from (16) and (31). In Line 3, the recursive function decode_channel with input parameters (nₑ = 2, i = 1) uses the probabilities in the array P₂ to obtain the decoding results of the intermediate vector (X₁^{(2)}, X₂^{(2)}, ..., X_n^{(2)}). The parameters here refer to the first traversed adjacent-bits-channel V_{2}^{(2)}. As indicated in (32), the decoding results (\hat{x}_1^{(2)}, \hat{x}_2^{(2)}, ..., \hat{x}_n^{(2)}) are stored in B₂. The last step is to obtain (\hat{x}_1, \hat{x}_2, ..., \hat{x}_n) from (\hat{x}_1^{(2)}, \hat{x}_2^{(2)}, ..., \hat{x}_n^{(2)}) in Lines 4-6, where the calculation in Lines 5-6 follows from (32) and

\[ (X₁, X₂, ..., Xₙ) = (X₁^{(2)}, X₂^{(2)}, ..., Xₙ^{(2)})(G₂^{\text{polar}} ⊗ I_{n/2}). \]

Next we explain how the function decode_channel in Algorithm 4 works. For each 1 ≤ i ≤ nₑ - 1, we define a subarray Pₙₑ[i] of Pₙₑ as

\[ Pₙₑ[i] = (Pₙₑ[i, β, a, b], \beta \in \{1, 2, ..., nₑ \}, a ∈ \{0, 1\}, b ∈ \{0, 1\}). \quad (33) \]

In the whole decoding algorithm, we always calculate all the entries in Pₙₑ[i] before we call the function decode_channel with input parameters nₑ and i. For example, in Algorithm 3, we calculate the array P₂[1] = P₂ in Lines 1-2 before calling the function decode_channel(2, 1) in Line 3; in Algorithm 7, we calculate P₂nₑ[2i] in Lines 4-5 before calling decode_channel(2nₑ/2i) in Line 5; in Algorithm 9, we first calculate P₂nₑ[2i - 1] in Lines 2-3 and then call decode_channel(2nₑ/2i - 1) in Line 4. If 1 ≤ i ≤ nₑ - 2, decode_channel(nₑ, i) uses Pₙₑ[i] to decode X_{β+i-1}/nₑ in Lines 5-6, 1 ≤ β ≤ nₑ/nₑ and stores the decoding result \( \hat{x}_{β+i-1}/nₑ \) in Bₙₑ[i, β]. If \( i = nₑ - 1 \), decode_channel(nₑ, i) uses Pₙₑ[i - 1] to decode X_{β+i-1}/nₑ in Lines 5-6, 1 ≤ β ≤ nₑ/nₑ and stores the decoding result \( \hat{x}_{β+i-1}/nₑ \) in Bₙₑ[nₑ - 1, β].

The implementation of decode_channel is divided into four cases. The first case nₑ = n is the boundary case, where we can directly decode Uᵢ (and Uᵢ+1 if i = nₑ - 1) from Pₙₑ[i]; see Algorithm 5. Note that we do not utilize the frozen value Uᵢ+1 when decoding the message bit Uᵢ for i < nₑ - 1. This decision is based on the observation that it does not lead to any significant differences in the decoding performance. In the other three cases, we decode X_{β+i-1}/nₑ and store the decoding result \( x_{β+i-1}/nₑ \) in Bₙₑ[nₑ - i, β].

Below we explain these three cases separately.

By Lemma 3, 2i ∈ I₂ⁿₑ⁻¹ implies that V_{2i}^{(2nₑ)} = (V_{i}^{(nₑ)})₂ ⊗ V_{2i}^{(2nₑ)}, which means that \( \hat{x}_{i}^{(nₑ)} \) in this case, decode_channel calls the function decode_swapped_channel in Algorithm 7. The first step in Algorithm 7 is to calculate P₂nₑ[2i - 1], P₂nₑ[2i] and P₂nₑ[2i + 1] from Pₙₑ[i] according to the above recursive relation; see Lines 2, 5, 8. Note that we encapsulate the calculation of transition probabilities for adjacent-bits-channels within the function calculate_probabilities, see Algorithm 6 for detail.
Algorithm 5 decode_boundary_channel(i)

Input: index i in the last layer (n_e = n)
1 if i ≤ n - 2 then
  // Only decode U_i
  \[ \text{B}_n[i, 1] \leftarrow \arg \max_{a \in \{0, 1\}} \sum_{b \in \{0, 1\}} P_n[i, 1][a, b] \]
  \( \triangleright U_i \) is an information bit
2 else
3  \[ \text{B}_n[i, 1] \leftarrow \text{frozen value of } U_i \]
4  \( \triangleright U_i \) is a frozen bit
5 else
6  \( \triangleright \) Decode both U_{n-1} and U_n.
7 if n - 1, n \not\in \mathcal{A} then
8  \[ \begin{align*}
9  & \text{B}_n[n - 1, 1], \text{B}_n[n, 1] \leftarrow \\
10  & \text{frozen bits } (U_{n-1}, U_n) \\
11  & \triangleright U_{n-1} \text{ and } U_n \text{ are both frozen bits}
12 \end{align*} \]
13 else if n - 1 \not\in \mathcal{A}, n \in \mathcal{A} then
14  \[ \begin{align*}
15  & \text{B}_n[n - 1, 1] \leftarrow \text{frozen value of } U_n \\
16  & \text{B}_n[n, 1] \leftarrow \arg \max_{a \in \{0, 1\}} P_n[n - 1, 1][a, U_n] \\
17  & \triangleright \text{information bit } U_{n-1}, \text{frozen bit } U_n
18 \end{align*} \]
19 else if n - 1 \in \mathcal{A}, n \not\in \mathcal{A} then
20  \[ \begin{align*}
21  & \text{B}_n[n - 1, 1] \leftarrow \text{frozen value of } U_{n-1} \\
22  & \text{B}_n[n, 1] \leftarrow \arg \max_{b \in \{0, 1\}} P_n[n - 1, 1][U_{n-1}, b] \\
23  & \triangleright \text{frozen bit } U_{n-1}, \text{information bit } U_n
24 \end{align*} \]
25 else
26  \[ \begin{align*}
27  & \text{B}_n[n - 1, 1], \text{B}_n[n, 1] \leftarrow \\
28  & \arg \max_{r, a \in \{0, 1\}} P_n[n - 1, 1][a, b] \\
29  & \triangleright U_{n-1} \text{ and } U_n \text{ are both information bits}
30 \end{align*} \]
31 return

In Line 3, decode_channel(2n_c, 2i - 1) uses P_{2n_c}[2i - 1] to decode \( X_{\beta + (2i - 1) \cdot n/(2n_c)} \), \( 1 \leq \beta \leq n/(2n_c) \) and stores the decoding results in B_{2n_c}[2i - 1, 1], \( 1 \leq \beta \leq n/(2n_c) \). Similarly, in Line 6, decode_channel(2n_c, 2i) uses P_{2n_c}[2i] to decode \( X_{\beta + (2i) \cdot n/(2n_c)} \), \( 1 \leq \beta \leq n/(2n_c) \) and stores the decoding results in B_{2n_c}[2i, 1], \( 1 \leq \beta \leq n/(2n_c) \). If \( i \leq n_c - 2 \), then decode_channel(2n_c, 2i + 1) in Line 9 only decodes one bit \( X_{\beta + (2i + 1) \cdot n/(2n_c)} \) for each \( \beta \in \{1, 2, \ldots, n/(2n_c)\} \); if \( i = n_c - 1 \) (i.e., \( 2i + 1 = 2n_c - 1 \)), then decode_channel(2n_c, 2i + 1) decodes two bits \( X_{\beta + (2i+1) \cdot n/(2n_c)} \). In Line 3 and Line 6, we only decode one bit for each value \( \beta \) because \( 2i - 1 < 2i \leq 2n_c - 2 \) for all \( 1 \leq i \leq n_c - 1 \). To summarize, after executing the first 9 lines of Algorithm 7, we have the following decoding results stored in the array B_{2n_c}: When \( i \leq n_c - 2 \), we have

\[
\begin{align*}
\text{B}_{2n_c}[2i - 1, 1] &= \hat{x}_{\beta + (2i - 2) \cdot n/(2n_c)}, \\
\text{B}_{2n_c}[2i, 1] &= \hat{x}_{\beta + (2i - 1) \cdot n/(2n_c)}, \\
\text{B}_{2n_c}[2i + 1, 1] &= \hat{x}_{\beta + (2i) \cdot n/(2n_c)}, \\
\text{B}_{2n_c}[2i + 2, 1] &= \hat{x}_{\beta + (2i + 1) \cdot n/(2n_c)}, \\
\end{align*}
\]
for \( 1 \leq \beta \leq n/(2n_c) \). (34)
In the latter case, we use the quantities in (34) to calculate $X_{\beta + (i-1)n/n_c}$, $1 \leq \beta \leq n/n_c$ and store the results in $B_{n_c}[i, \beta]$, $1 \leq \beta \leq n/n_c$; see Lines 10-15. In the latter case, we use the quantities in (35) to calculate $X_{\beta + (i-2)n/n_c}$, $X_{\beta + (i-1)n/n_c}$, $1 \leq \beta \leq n/n_c$ and store the results in $B_{n_c}[i-1, \beta], B_{n_c}[i-\beta], 1 \leq \beta \leq n/n_c$; see Lines 16-23. In Fig. 6(a), we further explain the calculations in Lines 13-15 and Lines 19-23. The structure of Algorithm 8 is exactly the same as that of Algorithm 7. The only difference is that we call decode_added_channel in Algorithm 4 when $2i \in \mathcal{T}_S^{(2n_c)}$. In this case, we have $V_{2i-1}^{(2n_c)} = (V_{2i-1}^{(n_c)})^\Delta, V_{2i}^{(2n_c)} = (V_{2i}^{(n_c)})^\Delta, V_{2i+1}^{(2n_c)} = (V_{2i+1}^{(n_c)})^\Delta$, and all the calculations in Algorithm 8 follow from this recursive relation.

The structure of Algorithm 9 differs from that of Algorithm 7 and Algorithm 8 in two places. First, Algorithm 9 only calculates $P_{2n_c}[2i-1]$ and calls decode_channel$(2n_c, 2i-1)$ when $2(i-1) \notin \mathcal{T}_S^{(2n_c)}$; see Lines 1-4. In contrast, Algorithm 7 and Algorithm 8 always calculate $P_{2n_c}[2i-1]$ and call decode_channel$(2n_c, 2i-1)$ to decode $X_{\beta + (2i-1)n/(2n_c)}, 1 \leq \beta \leq n/(2n_c)$ for all values of $i$. This is because we have already decoded $X_{\beta + (2i-2)n/(2n_c)}, 1 \leq \beta \leq n/(2n_c)$ when $2(i-1) \in \mathcal{T}_S^{(2n_c)}$, and this condition can only hold for the input parameters $n_c$ and $i$ in Algorithm 9. In both Algorithm 7 and Algorithm 8, we have $2i \in \mathcal{T}_S^{(2n_c)}$, and the fully separated requirement (5) implies that $2(i-1) \notin \mathcal{T}_S^{(2n_c)}$. Second, Algorithm 9 only calculates $P_{2n_c}[2i+1]$ and calls decode_channel$(2n_c, 2i+1)$ when $i = n_c - 1$; see Lines 19-21. In contrast, Algorithm 7 and Algorithm 8 calculate $P_{2n_c}[2i+1]$ and call decode_channel$(2n_c, 2i+1)$ to decode $X_{\beta + 2n_c/(2n_c)}, 1 \leq \beta \leq n/(2n_c)$ for all $1 \leq i \leq n_c - 1$. This is because $X_{\beta + 2n_c/(2n_c)}, 1 \leq \beta \leq n/(2n_c)$ is needed in the calculation of $X_{\beta + (i-1)n/(2n_c)}, 1 \leq \beta \leq n/n_c$ if and only if $2i \in \mathcal{T}_S^{(2n_c)}$; see Fig. 6 and Fig. 7.

The calculations in Lines 14-16 and Lines 23-27 of Algorithm 9 are explained in Fig. 7. More specifically, Fig. 7(a) and Fig. 7(c) tell us that when $2(i-1) \notin \mathcal{T}_S^{(2n_c)}$,
Algorithm 8 decode_added_channel(nc, i)
Input: nc ∈ {2, 4, 8, ..., n} and index i satisfying 2i ∈ I_S(nc)
1 for β ∈ {1, 2, ..., n/(2nc)}, a ∈ {0, 1} and b ∈ {0, 1} do
2   P2nc[2i − 1, β][a, b] ← calculate_probability(nc, i, β, ∇, a, b)
3 decode_channel(2nc, 2i − 1)
4 for β ∈ {1, 2, ..., n/(2nc)}, a ∈ {0, 1} and b ∈ {0, 1} do
5   P2nc[2i, β][a, b] ← calculate_probability(nc, i, β, ∇, a, b)
6 decode_channel(2nc, 2i)
7 for β ∈ {1, 2, ..., n/(2nc)}, a ∈ {0, 1} and b ∈ {0, 1} do
8   P2nc[2i + 1, β][a, b] ← calculate_probability(nc, i, β, ∇, a, b)
9 decode_channel(2nc, 2i + 1)
10 if i ≤ nc − 2 then
11   > Only decode one bit X\(\beta(n_{c})\) for each β
12   for β ∈ {1, 2, ..., n/(2nc)} do
13      β′ ← β + n/(2nc)
14      B2nc[i, β] ← B2nc[2i − 1, β] + B2nc[2i, β] + B2nc[2i + 1, β]
15      B2nc[i, β] ← B2nc[2i, β] + B2nc[2i + 1, β]
16   > See Fig. 6(b) for an explanation
17 else
18   > Decode two bits X\(\beta(n_{c})\), X\(\beta(n_{c}−1)n/n_{c}\) for each β
19   for β ∈ {1, 2, ..., n/(2nc)} do
20      β′ ← β + n/(2nc)
21      B2nc[i, β] ← B2nc[2i, β] + B2nc[2i − 1, β]
22      B2nc[i, β] ← B2nc[2i, β] + B2nc[2i − 1, β] + B2nc[2i + 1, β]
23   > See Fig. 6(b) for an explanation
24 return

we have the following relation
\[
X^{\beta(n_{c})}_{\beta+(i-1)n/n_{c}} = X^{2n_{c}}_{\beta+(2i-2)n/2n_{c}} + X^{2n_{c}}_{\beta+(2i-1)n/2n_{c}},
\]
\[
X^{\beta(n_{c})}_{\beta+(i-1)n/n_{c}} = X^{2n_{c}}_{\beta+(2i-2)n/2n_{c}} + X^{2n_{c}}_{\beta+(2i-1)n/2n_{c}},
\]
\[
X^{\beta+i/n_{c}}_{\beta+2i+1/n_{c}} = X^{2n_{c}}_{\beta+2i+1/n/2n_{c}} + X^{2n_{c}}_{\beta+2i+1/n/2n_{c}},
\]
\[
X^{\beta+i/n_{c}}_{\beta+2i+1/n_{c}} = X^{2n_{c}}_{\beta+2i+1/n/2n_{c}} + X^{2n_{c}}_{\beta+2i+1/n/2n_{c}},
\]
\[
(36)
\]
where \(\beta′ = \beta + n/n_{c}\). The relation for the case 2(i − 1) ∈ I_S(nc) is obtained from replacing \(X^{2n_{c}}_{\beta+(2i-2)n/2n_{c}}\) with \(X_{\beta+(2i-3)n/2n_{c}}\) in the first equation above. In this case, we move the decoding result \(X^{2n_{c}}_{\beta+(2i-3)n/2n_{c}}\) stored in \(B2nc[2i-2, \beta]\) to \(B2nc[2i-1, \beta]\); see Lines 5-7 in Algorithm 9. Then the calculations in Lines 14-16 and Lines 23-27 follow from (36) and (32).

In Fig. 8, we use the (16, 8) ABS+ polar code defined in Fig. 5 as a concrete example to illustrate the recursive structure of the function decode_channel in Algorithm 4.

In the whole decoding procedure, we call the function decode_channel(nc, i) exactly once for each \(nc \in \{2, 4, 8, ..., n\}\) and each \(1 \leq i \leq nc - 1\); see Fig. 8 for an illustration. It is easy to see that the time complexity\(^3\) of decode_channel(nc, i) is \(O(n_{c})\). Since \(i\) takes

\(^3\)We do not include the running time of the recursive calls decode_channel(2nc, 2i − 1), decode_channel(2nc, 2i) and decode_channel(2nc, 2i + 1) in the time complexity of decode_channel(nc, i).
$n_c - 1$ values and $n_c$ takes $\log(n)$ values, the time complexity of the SC decoder is $O(n/n_c) \cdot n_c - \log(n) = O(n \log(n))$.

**Proposition 2:** The time complexity of the SC decoder for ABS+ polar codes is $O(n \log(n))$.

### A. Space-Efficient Version of the SC Decoder

As mentioned earlier, we can reduce the space complexity of the SC decoder from $O(n \log(n))$ to $O(n)$. As we can see from Algorithms 4–8, we only use the entries stored in $P_{n_c}[i]$ when we call the function `decode_channel` with input parameters $(n_c, i)$. The entries in $P_{n_c}[i]$ are never used again in the whole decoding algorithm after the function `decode_channel(n_c, i)` returns. Moreover, for $1 \leq i \leq n_c - 2$, the function `decode_channel(n_c, i + 1)` is called after `decode_channel(n_c, i)` returns. Therefore, for each $n_c \in \{2, 4, 8, \ldots, n\}$, we can reduce the 4-dimensional array

$$
(P_{n_c}[i, \beta][a, b], 1 \leq i \leq n_c - 1, 1 \leq \beta \leq n/n_c, a \in \{0, 1\}, b \in \{0, 1\})
$$

to a 3-dimensional array

$$
(P_{n_c}[\beta][a, b], 1 \leq \beta \leq n/n_c, a \in \{0, 1\}, b \in \{0, 1\})
$$

dropping the index $i$. Each entry $P_{n_c}[\beta][a, b]$ stores $V_{i, \beta}[x^{(n_c)}(i), y^{(n_c)}(i)][a, b]$, when we call the function `decode_channel` with input parameters $(n_c, i)$.

Upon completion of the function `decode_channel(n_c, i)`, we note that (1) If $i \in \mathcal{I}^{(n_c)}$, the decoding results of $X^{(n_c)}(i + 1)/n_c)$ will be used to compute the transition probabilities for adjacent-bits-channel $V_{i + 1}^{(n_c)}$ (see Fig. 7 (b)); (2) If $i - 1 \in \mathcal{I}^{(n_c)}$, the decoding results of $X^{(n_c)}(i + 1)/n_c)$ will be used to compute the transition probabilities for adjacent-bits-channel $V_{i + 1}^{(n_c)}$ (see Fig. 7 (c)). For these two exceptional cases, we need an additional helper space $H$ to store these intermediate results. More precisely, we reduce the 2-dimensional array $(B_{n_c}[i, \beta], 1 \leq i \leq n_c, 1 \leq \beta \leq n/n_c)$ to a 1-dimensional array $(B_{n_c}[\beta], 1 \leq \beta \leq n/n_c)$ in addition, we introduce a new 1-dimensional array $(H_{n_c}[\beta], 1 \leq \beta \leq n/n_c)$. Each entry $H_{n_c}[\beta]$ stores $\hat{x}_{\beta + (i - 1)(n)/n_c}$ when the function `decode_channel(n_c, i)` returns, and each entry $H_{n_c}[\beta]$ stores $\hat{x}_{\beta + (n_c - 1)(n)/n_c}$ when the function `decode_channel(n_c, n_c - 1)` returns.

To summarize, we have three data structures

$$(P_{n_c}[\beta][a, b], 1 \leq \beta \leq n/n_c, a \in \{0, 1\}, b \in \{0, 1\}), \quad (B_{n_c}[\beta], 1 \leq \beta \leq n/n_c), \quad (H_{n_c}[\beta], 1 \leq \beta \leq n/n_c),$$

in the space-efficient version of the SC decoder. For each $n_c \in \{2, 4, 8, \ldots, n\}$, the arrays $P_{n_c}$, $B_{n_c}$ and $H_{n_c}$ have $6n/n_c$ entries in total. Therefore the space complexity of the space-efficient SC decoder is $6(n/2 + n/4 + n/8 + \cdots + n/n) = O(n)$.

**Proposition 3:** The space complexity of the space-efficient SC decoder for ABS+ polar codes is $O(n)$.

Next we show how to modify Algorithms 3–8 to the space-efficient version. Algorithms 3–6 only require a few modifications, which are listed below.

(i) In Algorithm 3, Algorithm 5 and Algorithm 6, we replace $P_{n_c}[\beta][a, b]$ with $P_{n_c}[\beta][a, b]$ and replace $B_{n_c}[i, \beta]$ with $B_{n_c}[\beta]$ for all $n_c \in \{2, 4, 8, \ldots, n\}$, $1 \leq i \leq n_c - 1, 1 \leq \beta \leq n/n_c$ and $a, b \in \{0, 1\}$.

(ii) In Algorithm 3 and Algorithm 5, we replace $B_{n_c}[n_c, \beta]$ with $H_{n_c}[\beta]$ for all $n_c \in \{2, 4, 8, \ldots, n\}$ and $1 \leq \beta \leq n/n_c$.

(iii) After Line 8 in Algorithm 4, we add the following three lines

\begin{center}
\begin{verbatim}
if $i \in \mathcal{I}^{(n_c)}$ or $i - 1 \in \mathcal{I}^{(n_c)}$ then
  for $\beta \in \{1, 2, \ldots, n/n_c\}$ do
    $H_{n_c}[\beta] \leftarrow B_{n_c}[\beta]
\end{verbatim}
\end{center}

(iv) In Algorithm 6, we replace Line 12 with

\begin{center}
\begin{verbatim}
r_1 \leftarrow B_{n_c}[\beta], \quad r_2 \leftarrow a, \quad r_3 \leftarrow b
and replace Line 21 with
  r_1 \leftarrow B_{n_c}[\beta], \quad r_2 \leftarrow B_{n_c}[\beta], \quad r_3 \leftarrow a, \quad r_4 \leftarrow b
\end{verbatim}
\end{center}
Fig. 9. Performance of ABS+ polar codes, ABS polar codes, and standard polar codes over the binary-input AWGN channel. The legend “ST” refers to standard polar codes. The CRC length is chosen from the set \{4, 8, 12, 16, 20\} to minimize the decoding error probability. The parameter $L$ is the list size.

For standard polar codes, we always choose $L = 32$. For ABS+ and ABS polar codes, we test two different list sizes $L = 20$ and $L = 32$.

Algorithms 7-9 require more changes than Algorithms 3-6. In Algorithm 10, we present the space-efficient version of Algorithm 7. In Algorithm 11, we present the space-efficient version of Algorithm 8. Finally, we present the space-efficient version of Algorithm 9 in Algorithm 12. Algorithms 12-11 are given in Appendix VI.
VI. SIMULATION RESULTS

We conduct extensive simulations over binary-input AWGN channels to compare the performance of ABS+ polar codes, ABS polar codes, and standard polar codes. We run simulations for 4 different choices of code length 256, 512, 1024, 2048. For each choice of code length, we test 3 different
code rates 0.3, 0.5, and 0.7. The comparison of decoding error probability is given in Fig. 9 and Fig. 10. Specifically, Fig. 9 contains the plots for code length 256 and 512; Fig. 10 contains the plots for code length 1024 and 2048. The comparison of decoding time is given in Table I.

In Fig. 9–10 and Table I, for each choice of code length and code dimension, we compare the performance of the following 5 decoders. (1) ST, \( L = 32 \): SCL decoder for standard polar codes with list size 32 and optimal CRC length; (2) ABS, \( L = 20 \): SCL decoder for ABS polar codes with list size 20 and optimal CRC length; (3) ABS, \( L = 32 \): SCL decoder for ABS polar codes with list size 32 and optimal CRC length; (4) ABS+, \( L = 20 \): SCL decoder for ABS+ polar codes with list size 20 and optimal CRC length; (5) ABS+, \( L = 32 \): SCL decoder for ABS+ polar codes with list size 32 and optimal CRC length. The optimal CRC length is chosen from the set \{4, 8, 12, 16, 20\} to minimize the decoding error probability.

From Table I we can see that decoders (1),(2),(4) have more or less the same running time; decoders (3),(5) have more or less the same running time. Moreover, the decoding time of decoders (3),(5) is longer than that of decoders (1),(2),(4) by roughly 60%. Note that if we set the list size in the SCL decoder to be the same, then the decoding time of ABS+ polar codes and ABS polar codes is very close to each other.

As for the decoding error probability, we mainly compare the performance of decoders (1), (2), and (4) because they have similar decoding time. As we can see from Fig. 9–10, ABS+ polar codes with list size 20 improves upon ABS polar codes.
Algorithm 11 decode_added_channel(n_c, i)
Space-Efficient Version

Input: \( n_c \in \{2, 4, 8, \ldots, n\} \) and index \( i \) satisfying
\[ \beta \in \mathbb{Z}^{2n_c}(2) \]
1 for \( \beta \in \{1, 2, \ldots, n/(2n_c)\}, a \in \{0, 1\} \) and \( b \in \{0, 1\} \) do
2 \[ \begin{aligned}
&\text{P}_{2n_c}[^{\beta}][a, b] \leftarrow \\
&\text{calculate_probability}(n_c, i, \beta, \nabla, a, b)
\end{aligned} \]
3 decode_channel(2n_c, 2\( i \))
4 for \( \beta \in \{1, 2, \ldots, n/(2n_c)\} \) do
5 \[ B_{n_c}[^{\beta}] \leftarrow B_{2n_c}[^{\beta}] \]
6 \[ B_{n_c}[^{\beta}] \leftarrow B_{2n_c}[^{\beta}] \]
7 if \( \beta \in \{1, 2, \ldots, n/(2n_c)\}, a \in \{0, 1\} \) and \( b \in \{0, 1\} \) do
8 \[ \begin{aligned}
&\text{P}_{n_c}[^{\beta}][a, b] \leftarrow \\
&\text{calculate_probability}(n_c, i, \beta, \nabla, a, b)
\end{aligned} \]
9 decode_channel(2n_c, 2\( i \))
10 for \( \beta \in \{1, 2, \ldots, n/(2n_c)\} \) do
11 \[ B_{n_c}[^{\beta}] \leftarrow B_{n_c}[^{\beta}] \]
12 \[ B_{n_c}[^{\beta}] \leftarrow B_{n_c}[^{\beta}] + B_{2n_c}[^{\beta}] \]
13 \[ B_{n_c}[^{\beta}] \leftarrow B_{2n_c}[^{\beta}] + B_{2n_c}[^{\beta}] \]
14 \[ B_{n_c}[^{\beta}] \leftarrow B_{2n_c}[^{\beta}] \]
15 if \( i \leq n_c - 2 \) then
16 \[ \begin{aligned}
&\text{Only one decode one bit } X_{\beta(i-1)n/n_c} \text{ for each } \\
&\beta \in \{1, 2, \ldots, n/(2n_c)\} \text{ do}
\end{aligned} \]
17 \[ \beta \leftarrow \beta + n/(2n_c) \]
18 \[ B_{n_c}[\beta] \leftarrow B_{n_c}[^{\beta}][\beta] + B_{2n_c}[\beta] \]
19 \[ B_{n_c}[\beta] \leftarrow B_{n_c}[\beta] + B_{2n_c}[\beta] \]
20 \[ \text{See Fig. 6(b) for an explanation} \]
21 else
22 \[ \begin{aligned}
&\text{Decode two bits } X_{\beta(i-2)n/n_c} \text{ for each } \\
&\beta \in \{1, 2, \ldots, n/(2n_c)\} \text{ do}
\end{aligned} \]
23 \[ \beta \leftarrow \beta + n/(2n_c) \]
24 \[ B_{n_c}[\beta] \leftarrow B_{n_c}[\beta] + B_{2n_c}[\beta] \]
25 \[ B_{n_c}[\beta] \leftarrow B_{2n_c}[\beta] + B_{2n_c}[\beta] \]
26 \[ B_{n_c}[\beta] \leftarrow B_{2n_c}[\beta] \]
27 \[ \text{See Fig. 6(b) for an explanation} \]
28 return

Algorithm 12 decode_original_channel(n_c, i)
Space-Efficient Version

Input: \( n_c \in \{2, 4, 8, \ldots, n\} \) and index \( i \) satisfying
\[ 1 \leq i \leq n_c - 1 \text{ and } 2i \not\in \mathbb{Z}^{2n_c}(2) \]
1 if \( i = 1 \) or \( 2(i-1) \not\in \mathbb{Z}^{2n_c}(2) \) then
2 \[ \begin{aligned}
&\text{Only one decode one bit } X_{\beta(i-1)n/n_c} \text{ for each } \\
&\beta \in \{1, 2, \ldots, n/(2n_c)\} \text{ do}
\end{aligned} \]
3 \[ \beta \leftarrow \beta + n/(2n_c) \]
4 \[ B_{n_c}[\beta] \leftarrow B_{n_c}[\beta] \]
5 \[ B_{n_c}[\beta] \leftarrow B_{n_c}[\beta] + B_{2n_c}[\beta] \]
6 \[ B_{n_c}[\beta] \leftarrow B_{2n_c}[\beta] + B_{2n_c}[\beta] \]
7 \[ B_{n_c}[\beta] \leftarrow B_{2n_c}[\beta] \]
8 \[ \text{See Fig. 6(b) for an explanation} \]
9 else
10 \[ \begin{aligned}
&\text{Decode two bits } X_{\beta(i-2)n/n_c} \text{ for each } \\
&\beta \in \{1, 2, \ldots, n/(2n_c)\} \text{ do}
\end{aligned} \]
11 \[ \beta \leftarrow \beta + n/(2n_c) \]
12 \[ B_{n_c}[\beta] \leftarrow B_{n_c}[\beta] + B_{2n_c}[\beta] \]
13 \[ B_{n_c}[\beta] \leftarrow B_{2n_c}[\beta] + B_{2n_c}[\beta] \]
14 \[ B_{n_c}[\beta] \leftarrow B_{2n_c}[\beta] \]
15 \[ \text{See Fig. 6(b) for an explanation} \]
16 return

with list size 20 by 0.1 dB–0.25 dB; ABS+ polar codes with list size 20 improves upon standard polar codes with list size 32 by 0.15 dB–0.35 dB. For instance, in Fig. 10(e) for the case of length \( n = 2048 \) and dimension \( k = 1024 \), we can observe that when the word error rate (WER) is \( 10^{-7} \) and the list size is 20, the signal-to-noise ratio requirement for ABS+ polar code is approximately 0.25 dB lower than that of the ABS polar code. Moreover, in Fig. 9(b) for the case of length \( n = \)
512 and dimension $k = 154$, we can see that when the WER is $2 \times 10^{-5}$, the signal-to-noise ratio requirement for ABS+ polar code with list size 20 is approximately 0.35 dB lower than that of the standard polar code with list size 32. Finally, if we set the list size to be 32 for both ABS+ and standard polar codes, then ABS+ polar codes demonstrate 0.2 dB–0.45 dB improvement over standard polar codes.

As a final remark, the implementations of all the algorithms in this paper are available at the website https://github.com/PlumJelly/ABS-Polar

APPENDIX

SPACE-EFFICIENT DECODING ALGORITHMS

See Algorithms 10–12.

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