BER-Based Physical Layer Security with Finite Codelength: Combining Strong Converse and Error Amplification

Il-Min Kim, Byoung-Hoon Kim, and Joon Kui Ahn

Department of Electrical and Computer Engineering
Queen’s University
Kingston, ON, Canada K7M 2A8
E-mail: ilmin.kim@queensu.ca

January 4, 2015
Abstract

A bit error rate (BER)-based physical layer security approach is proposed for finite blocklength. For secure communication in the sense of high BER, the information-theoretic strong converse is combined with cryptographic error amplification achieved by substitution permutation networks (SPNs) based on confusion and diffusion. For discrete memoryless channels (DMCs), an analytical framework is provided showing the tradeoffs among finite blocklength, maximum/minimum possible transmission rates, and BER requirements for the legitimate receiver and the eavesdropper. Also, the security gap is analytically studied for Gaussian channels and the concept is extended to other DMCs including binary symmetric channels (BSCs) and binary erasure channels (BECs). For fading channels, the transmit power is optimized to minimize the outage probability of the legitimate receiver subject to a BER threshold for the eavesdropper.

Index Terms

BER, error amplification, finite blocklength, physical layer security, strong converse.

I. INTRODUCTION

Security is a critical issue in communications [1] and it is particularly challenging with a growing number of different wireless communication applications and various wireless devices. Due to the broadcast nature of wireless medium, the wireless security is inherently more vulnerable than the wired security: the eavesdropper may overhear and interpret the messages in wireless communications more easily than in wireline communications. Traditionally, the issue of security has been addressed at a higher layer by cryptography, which requires secret keys. A problem of this approach is that it is often challenging to distribute and manage the secret keys, especially for many emerging wireless networks. Furthermore, once the devices are physically compromised by an adversary, the communication is no longer secure.

As a fundamentally different approach, the physical layer security, particularly information theoretic security, has received a lot of attention. The information-theoretic security is based on the pioneering work of [2], where the channel from the transmitter (Alice) to the eavesdropper (Eve) was assumed to be a degraded version of the channel from Alice to the legitimate receiver (Bob), namely the degraded wiretap channel. For this channel, Wyner derived the capacity-equivocation region. Later, this work was extended to the non-degraded case, where the eavesdropper’s channel is not necessarily a degraded version of the legitimate user’s channel [3], and also applied to Gaussian channels [4]. Recently, the information theoretic security and/or
the physical layer security have regained much interest for secure wireless communications.

In most of the works in the area of physical layer security, the security metric is defined based on mutual information between Alice and Eve. If the mutual information is strictly zero, it is perfectly secure, called perfect security [1]. With perfect security, Eve cannot obtain any additional information about Alice’s message from Eve’s received signal. However, in order to ensure such perfect security, the entropy of the secret key must not be smaller than the entropy of the source message. In real communications, therefore, it is not practical to try to achieve the perfect security. Addressing this issue, two non-perfect security notions have been extensively studied: weak secrecy [2] and strong secrecy. The weak secrecy requires that the mutual information rate, i.e., mutual information divided by the blocklength (or codelength), approaches zero when the blocklength goes to infinity. On the other hand, the strong secrecy requires that the mutual information itself approaches zero when the blocklength goes to infinity. Many researchers have designed codes providing the weak secrecy or strong secrecy [5]–[10]. Vast majority of the works have been devoted to the weak secrecy, mostly based on low-density parity-check (LDPC) codes [5], [6] or polar codes [7] (also, see the references in [10]). Because designing codes to achieve the strong secrecy is generally much more difficult, the works for the strong secrecy were generally limited to simplistic scenarios such as noiseless Bob’s channel [7], [8] or binary symmetric channel [9]. However, it has been argued that the weak secrecy might be a too weak security condition [11], [12], and in fact, one can easily construct examples of codes achieving weak security that are never secure in practice [10]. A problem of those codes in [5]–[10] is that they are not directly applicable to continuous-input channels, such as additive white Gaussian noise (AWGN) or fading channels. Another (perhaps more serious) problem is that, for finite blocklength, it is not clear how to evaluate or quantify the strength of security actually achieved by the codes designed based on strong or weak secrecy secrecy. Unless the blocklength is very long, the codes might not be secure enough to be used in practical systems, especially for the case of weak secrecy.

Other than the information theoretic security notions based on mutual information, there are few other security measures considered in the literature. For example, the signal-to-interference-plus-noise ratio (SINR) has been used as a secrecy measure in the area of physical layer security based on signal processing techniques [13], [14]. However, it is unclear how to exactly set an SINR threshold and to evaluate what strength (or kind) of security can be actually achieved
Another security measure considered in the literature is the block error probability, i.e., decoding error probability of codeword. When the transmission rate is above the channel capacity, by the strong converse \[15\], the block error probability approaches one as the blocklength tends to infinity. Given a security condition in terms of block error probability, for a Gaussian wiretap channel, the authors of \[16\] studied the asymptotic transmission rate and the rate for finite blocklength using a rate approximation expression \[17\, \text{Theorem 54}\] for AWGN channels. In \[18\], a coset lattice code was designed to ensure high block error probability at Eve and low block error probability at Bob. However, a limitation of the approach based on the block error probability only is that high block error probability at Eve does not necessarily mean secure communication. This is because a block error event simply means that there is at least one bit error within a block (or codeword). As an example, if there is always only one bit error in a block, the block error probability is one. However, all the remaining bits except the particular single bit can be decoded by Eve, which is certainly not secure.

Arguably, a practically effective and useful security measure in the physical layer security might be the bit error rate (BER). If it is possible to ensure that Eve’s BER is (very close to) 0.5, she essentially cannot recover any information bits transmitted by Alice. In \[19\], for AWGN channels, punctured LDPC codes were designed to ensure high BER at Eve. The analysis was limited to asymptotic case of LDPC codes and Eve’s BER is evaluated only by simulations, from which it is not easy to obtain any theoretical insights. In \[20\], to induce high BER at Eve for AWGN channels, Bose-Chaudhuri-Hocuenghem (BCH) codes and LDPC codes are combined with scrambling/descrambling. The BER analysis of BCH codes was based on an approximate BER equation of \[21\] under the assumption of bounded-distance decoding with hard decision, and the study on LDPC codes was purely based on simulations.

In this paper, we also adopt the BER as the security measure for Eve. Using Gallager’s random coding exponent and the strong converse over general discrete memoryless channels (DMCs), we first ensure that Bob’s block error probability tends to zero and Eve’s block error probability tends to one. To amplify the errors such that Eve’s BER is close to 0.5, we then utilize substitution permutation networks (SPNs). In particular, the error amplification by SPN is not only mathematically analyzed based on the ideal modeling, but also numerically evaluated based on actual simulation of a real SPN. Given BER requirements for Bob and Eve, for finite blocklength, we analyze the maximum and minimum possible transmission rates. Also, the
security gap is defined and analyzed for AWGN channels and then the concept is extended to other DMCs. Focusing on Gaussian-input fading channels, we analytically optimize the transmit power to minimize Bob’s reliability outage probability, subject to a security condition given in terms of a BER lower-bound threshold for Eve. The summary of the contributions is as follows:

- For secure communication in the sense of high BER, the information-theoretic strong converse is combined with cryptographic error amplification achieved by SPNs based on confusion and diffusion.
- For DMCs, an analytical framework is provided showing the trade-offs among finite block-length, maximum/minimum possible transmission rates, and BER requirements for Bob and Eve.
- For Gaussian channels, with finite blocklength, the security gap is analytically studied and the concept is extended to other DMCs including binary symmetric channels (BSCs) and binary erasure channels (BECs).
- For fading channels, with finite blocklength, the transmit power is analytically optimized to minimize Bob’s outage probability subject to a BER threshold for Eve.

A practical benefit of the BER-based physical layer security is particularly evident when both Bob’s and Eve’s channels are good and the channel quality difference is small: \( C_b > C_e \gg 1 \) with \( C_b - C_e \ll 1 \), where \( C_b \) is Bob’s capacity and \( C_e \) is Eve’s capacity. If the weak secrecy or strong secrecy constraint is imposed, the transmission rate is bounded by the secrecy capacity given by \( C_b - C_e \ll 1 \) for the channels such as symmetric degraded wiretap channels \([22]\) or Gaussian channels \([4]\). On the other hand, if the high BER condition is imposed as a security constraint and our approach is taken, the transmission rate can go up to \( C_b \gg 1 \). Another benefit of the proposed approach is that, for finite blocklength, we can ensure a high target BER requirement for Eve, whereas for weak/strong secrecy, it is not entirely clear how to ensure a particular security requirement with finite blocklength.

The rest of this paper is organized as follows. In Section II, Gallager’s random coding exponent and the strong converse are reviewed to derive Bob’s block error probability upper-bound and Eve’s block error probability lower-bound. Also, it is demonstrated that the errors can be effectively amplified by SPNs. In Section III, we first combine the strong converse and the SPNs. Then the maximum/minimum rates and security gaps are analyzed given finite blocklength and the BER requirements for Bob and Eve. Also, for fading channels, the transmission power
is optimized to minimize the reliability outage probability subject to a security condition. In Section IV, some numerical results are presented and the paper is concluded in Section V.

Notation: We use $A := B$ to denote that $A$, by definition, is equal to $B$, and we use $A := B$ to denote that $B$, by definition, is equal to $A$. Also, $\mathcal{CN}(0, \sigma^2)$ denotes a circularly symmetric complex Gaussian distribution with variance $\sigma^2$ (or variance $\sigma^2/2$ per dimension).

II. GALLGER FUNCTION, STRONG CONVERSE, AND ERROR AMPLIFICATION

Assume that message $M$ represented by $K$ bits is transmitted by Alice. Using a code composed of $2^K$ codewords, the message is encoded into a codeword $X^n$ of $n$ symbols. The transmission rate $R$ is given by

$$R = \frac{K \ln 2}{n} \text{ (nats/channel use).} \quad (1)$$

Bob’s received codeword is denoted by $Y^n_b$ and Eve’s received codeword is denoted by $Y^n_e$. Assuming both channels are DMCs, they are described by the conditional probability distributions $f_{Y_b|X}(y_b|x)$ and $f_{Y_e|X}(y_e|x)$, respectively, for Bob and Eve. Let $\hat{M}_b$ and $\hat{M}_e$ denote the decoded messages at Bob and Eve, respectively. Let $C_b$ and $C_e$ denote the channel capacities for Bob and Eve, respectively.

A. Bob’s Block Error Probability based on Gallager Function

Let $C$ denote a code whose symbols $X$ are randomly generated by input distribution $q_X(x)$, which is simply denoted by $q(x)$ whenever there is no ambiguity. Let $P_{\text{err}}^b(R|C) = \Pr(M \neq \hat{M}_b|C)$ denote the decoding error probability of code $C$ at Bob. Let $P_{\text{err}}^b(R)$ denote the average probability over the ensemble of all codes at Bob. The ensemble average block error probability $P_{\text{err}}^b(R)$ at Bob can be upper-bounded as follows [23, Theorem 5.6.2]:

$$P_{\text{err}}^b(R) = \mathbb{E}[P_{\text{err}}^b(R|C)] \leq P_{\text{err}}^{b,U}(R, \rho, q(x)) \quad (2)$$

where the upper-bound $P_{\text{err}}^{b,U}(R, \rho, q(x))$ is given by

$$P_{\text{err}}^{b,U}(R, \rho, q(x)) = \exp\left(-n \left\{ E_0^b(\rho, q(x)) - \rho R \right\} \right), \quad 0 \leq \rho \leq 1. \quad (3)$$

In the above equation, Gallager function $E_0^b(\rho, q(x))$ is given by

$$E_0^b(\rho, q(x)) = -\ln \sum_{y_b} \left[ \sum_x q(x) f_{Y_b|X}(y_b|x)^{1+\rho} \right]^{1+\rho}, \quad 0 \leq \rho \leq 1. \quad (4)$$
where $\sum_x$ is replaced by $\int_x$ if $X$ is continuous, and $\sum_{Y_b}$ is replaced by $\int_{Y_b}$ if $Y_b$ is continuous. Since the upper-bound $P_{\text{err}}^{b,U}(R, \rho, q(x))$ is valid for any $0 \leq \rho \leq 1$ and for any distribution $q(x)$, the bound can be tightened by optimizing $\rho$ and $q(x)$ as follows:

$$\min_{0 \leq \rho \leq 1} \min_{q(x)} P_{\text{err}}^{b,U}(R, \rho, q(x))$$

(5)
or

$$\min_{0 \leq \rho \leq 1} \left\{ \max_{q(x)} E^b_0(\rho, q(x)) - \rho R \right\}.$$  

(6)

In this paper, we will use $\hat{q}(x)$ and $\hat{\rho}$ to denote the optimal distribution and optimal $\rho$, respectively, which are defined as follows:

$\hat{q}(x) = \arg \min_{q(x)} P_{\text{err}}^{b,U}(R, \rho, q(x)) = \arg \max_{q(x)} E^b_0(\rho, q(x))$  

(7)

$\hat{\rho} = \arg \min_{0 \leq \rho \leq 1} P_{\text{err}}^{b,U}(R, \rho, \hat{q}(x)) = \arg \max_{0 \leq \rho \leq 1} \left\{ E^b_0(\rho, \hat{q}(x)) - \rho R \right\}.$  

(8)

When $R < I_b(q(x))$, the exponent in (3) is positive with maximization over $\rho$ [23, Section 5.6, p. 143]:

$$\max_{0 \leq \rho \leq 1} \left\{ E^b_0(\rho, q(x)) - \rho R \right\} > 0, \quad R < I_b(q(x)).$$

(9)

When $R < C_b$, the exponent in (3) is positive with maximization over $q(x)$ and $\rho$ [23, Section 5.6, p. 143]:

$$\max_{0 \leq \rho \leq 1} \left\{ \max_{q(x)} E^b_0(\rho, q(x)) - \rho R \right\} > 0, \quad R < C_b.$$  

(10)

When $R < C_b$, therefore, there exists at least one code of which block error probability upper-bound tends exponentially to zero as $n \to \infty$. With the optimal $\hat{q}(x)$ yielding the tightest upper-bound, the asymptotic slope of $E^b_0(\rho, q(x))$ when $\rho$ approaches zero from the right is the capacity of Bob’s channel [23, Section 5.6]:

$$C_b = \lim_{\rho \to 0} \frac{1}{\rho} \max_{q(x)} E^b_0(\rho, q(x))$$

(11)

$$= \max_{q(x)} \left. \frac{\partial}{\partial \rho} E^b_0(\rho, q(x)) \right|_{\rho=0}.$$  

(12)
B. Eve’s Block Error Probability based on Arimoto’s Strong Converse

Let $P_{\text{err}}^e(R|C) = \Pr(M \neq \hat{M}_e|C)$ denote the block error probability of code $C$ at Eve. We first define $P_{\text{err}}^{\text{e,L}}(R, \rho', \bar{q}'(x))$ as follows

$$P_{\text{err}}^{\text{e,L}}(R, \rho', \bar{q}'(x)) = 1 - \exp(-n \{ E_0^e(\rho', \bar{q}'(x)) - \rho' R \}), \quad -1 < \rho' \leq 0 \tag{13}$$

where $E_0^e(\rho', \bar{q}'(x))$ is given by $\{4\}$ with $q(x)$, $f_{Y_b|x}(y_b|x)$, and $\rho$ replaced by $\bar{q}'(x)$, $f_{\hat{Y}_b|x}(\hat{y}_b|x)$, and $\rho'$, respectively. When $R > C_e$ and a priori probabilities are equal, Eve’s block error probability $P_{\text{err}}^e(R|C)$ of any code $C$ is be lower bounded by $[15]$, $[24]$ Eq. (3.9.21)):

$$P_{\text{err}}^e(R|C) \geq P_{\text{err}}^{\text{e,L}}(R, \rho', \bar{q}'(x)), \quad \forall C \tag{14}$$

where $\bar{q}'(x)$ is given by

$$\bar{q}'(x) = \arg\min_{\bar{q}'(x)} P_{\text{err}}^{\text{e,L}}(R, \rho', \bar{q}'(x)) = \arg\min_{\bar{q}'(x)} E_0^e(\rho', \bar{q}'(x)). \tag{15}$$

Note that, unlike the case of upper-bound, the single-letter expression $\{14\}$ of the lower-bound is obtained with the particular input distribution $\bar{q}'(x)$ $\{4\}$. Since lower-bound $P_{\text{err}}^{\text{e,L}}(R, \rho', \bar{q}'(x))$ is still valid for any $-1 < \rho' \leq 0$, the tightest bound can be obtained by optimizing $\rho'$ as follows:

$$\bar{\rho}' = \arg \max_{-1 < \rho' \leq 0} P_{\text{err}}^{\text{e,L}}(R, \rho', \bar{q}'(x)) = \arg \max_{-1 < \rho' \leq 0} \{ E_0^e(\rho', \bar{q}'(x)) - \rho' R \}. \tag{16}$$

When $R > C_e$, the exponent in $\{13\}$ is positive with maximization over $\rho'$ and minimization over $q'(x)$: $[15]$ Theorem 2] $[24]$ Theorem 3.9.1):

$$\max_{-1 < \rho' \leq 0} \left\{ \min_{q'(x)} E_0^e(\rho', q'(x)) - \rho' R \right\} > 0, \quad R > C_e. \tag{17}$$

When $R > C_e$ and a priori probabilities are equal, therefore, the error probability upper-bound of any code tends exponentially to one as $n \to \infty$. With the particular distribution $\bar{q}'(x)$ yielding the valid lower-bound for any code, the asymptotic slope of $E_0^e(\rho', \bar{q}'(x))$ when $\rho'$ approaches zero from the left is the capacity of Eve’s channel $[15]$:

$$C_e = \lim_{\rho' \to 0} \frac{1}{\rho'} \min_{\rho', \bar{q}'(x)} E_0^e(\rho', \bar{q}'(x)) \tag{18}$$

$$= \max_{\bar{q}'(x)} \left. \frac{\partial}{\partial \rho'} E_0^e(\rho', \bar{q}'(x)) \right|_{\rho' = 0}. \tag{19}$$

$^1$Compared to the upper-bound tightened by $\bar{q}(x)$, the lower-bound determined by $\bar{q}'(x)$ might be considered to be weaker or less tight because $\bar{q}'(x)$ is obtained by minimizing $E_0^e(\rho', q'(x))$ rather than maximizing it. In return, the obtained lower-bound is valid for all possible codes (rather than some codes in the ensemble as in the upper-bound case).
C. Confusion and Diffusion: Error Amplification by SPN in Cryptography

In this subsection, the issue of error amplification is discussed. In cryptography, error amplification has been extensively and systematically studied for various applications including hash functions and block ciphers such as Data Encryption Standard (DES) and Advanced Encryption Standard (AES) [25]. A most common approach is to use substitution-boxes (S-boxes), which are designed based on several criterions such as the completeness, avalanche property, etc. In particular, the avalanche property plays a very important role. This property was first introduced by Feistel [26]; but, the fundamental concept was actually based on Shannon’s confusion [1]. In [27], strict avalanche criterion (SAC) was defined as follows: SAC is satisfied if, whenever a single input bit is complemented, each of all output bits changes with a 50% probability. Also, high degree SAC can be defined [28]–[31]: SAC of degree \( l \) is satisfied if, whenever \( l \) input bits are complemented at the same time, each of the output bits changes with a 50% probability.

In general, it is very difficult to design large-size S-boxes satisfying SAC. In today’s practical cryptographic systems, therefore, small-size S-boxes are often used; for example, \( 8 \times 8 \) S-boxes are used for AES. In order to handle a larger number of input bits at the same time, substitution-permutation networks (SPNs) are often used. An SPN is composed of multiple parallel-connected S-boxes taking multiple input bits. The output bits from those S-boxes are permuted by a permutation box (P-box). Typically, an SPN is designed by implementing several rounds of alternating S-boxes and P-boxes. In fact, the design of alternating S-boxes and P-boxes is based on Shannon’s two fundamental security concepts: confusion and diffusion [1]. In SPNs for cryptographic applications, secret keys are typically used. In this paper, however, we do not use any secret keys for SPNs because we will use SPNs only to amplify the errors (rather than encrypting data as in cryptography). In the following, the error amplification effect of SPNs is evaluated first by analysis assuming ideal S-boxes and then by simulation using real S-boxes.

In [32], assuming ideal S-boxes satisfying SAC, the output error probability of the SPN was analyzed. Let \( K \) denote the number of input and output bits of the SPN. Let \( W_r \) denote the random variable representing the number of bit errors after round \( r \). Let \( B \) denote the number of input and output bits of each S-box. Assuming \( K \) is an integer multiples of \( B \), we use \( J = \frac{K}{B} \) to denote the number of S-boxes connected in parallel for each round. Let \( L_r \) denote

\(^3\)For example, AES has 10 rounds for 128 bit secret keys, 12 rounds for 192 bit secret keys, and 14 rounds for 256 bit secret keys.
the random variable representing the number of S-boxes in round $r$ affected by the bit errors. The distribution of $W_r$ is given by

$$q_{W_r}(w_r) = \sum_{l_r=1}^{J} f_{W_r|L_r}(w_r|l_r) \sum_{w_{r-1}=1}^{K} f_{L_r|W_{r-1}}(l_r|w_{r-1}) q_{W_{r-1}}(w_{r-1}), \text{ for } w_r = 1, \cdots, K$$

(20)

where

$$f_{L_r|W_{r-1}}(l_r|w_{r-1}) = \frac{A_1(l_r, w_{r-1})}{A_2(w_{r-1})}, \text{ for } l_r = 1, \cdots, J$$

(21)

$$A_1(l, w) = \sum_{i=J-l}^{J} (-1)^{i-(J-l)} \binom{i}{J-l} \binom{(J-i)B}{w}^+$$

(22)

$$A_2(w) = \binom{K}{w}$$

(23)

$$f_{W_r|L_r}(w_r|l_r) = \frac{1}{(2^B - 1)^{l_r}} \sum_{i=0}^{l_r} (-1)^i \binom{l_r}{i} \binom{(l_r-i)B}{w_r}^+$$

(24)

In the above equation, $\binom{a}{b}^+ = \binom{a}{b}$ if $a \geq b$; $\binom{a}{b}^+ = 0$ if $a < b$. Using $q_{W_r}(w_r)$, the BER at the output of the SPN after $r$ rounds can be determined as follows

$$P_{\text{BER}}^{\text{SPN}}(r, K) = \frac{1}{K} \sum_{w_r=1}^{K} w_r q_{W_r}(w_r), \quad r = 1, 2, \cdots .$$

(25)

In order to actually determine the BER using (25), the initial distribution $q_{W_0}(w_0)$ must be explicitly given. As an example, for the scenario where there is only a single input bit error, the initial distribution is given by

$$q_{W_0}(w_0) = \begin{cases} 1, & \text{if } w_0 = 1 \\ 0, & \text{otherwise} \end{cases}$$

(26)

In Fig. 1 the output BER analytically obtained by (20)–(26) is plotted for different sizes of SPNs with $B = 8$. The number $J$ of S-boxes for each round is given by $\frac{K}{8}$. One can see that, with a small number $r$ of rounds, the BER is generally smaller for larger $K$, because it takes more rounds for the case of large $K$ to spread the errors over the entire bits. However, for larger number of rounds (e.g., $r \geq 4$), the BER is essentially 0.5 regardless of the size $K$ of the SPN.

Above analysis and numerical results are based on the ideal S-boxes satisfying SAC. We now evaluate the BER of an actual SPN composed of real S-boxes. In this paper, as an example, we use the actual $8 \times 8$ S-boxes adopted for AES [25, Fig. 3.8], which is known to have good

\[\text{Although this expression is given in closed-form, it becomes difficult to use as } K \text{ increases, because the computational complexity grows with } K \text{ very quickly.}\]
avalanche property [33]. For the case of single input bit error, Fig. 2 shows the output BER obtained by simulations. One can see that, by increasing the number \( r \) of rounds, it is possible to make the output BER close to 0.5. This means that the input error can be effectively amplified by actual SPNs.

![Fig. 1. BER at the output of the SPN composed of 8 × 8 theoretical S-boxes satisfying SAC, when only one input bit is in error out of total \( K \) input bits. The number \( J \) of S-boxes for each round is given by \( \frac{K}{8} \). The BER is analytically obtained by (20)–(26).](image)

**III. SECURE TRANSMISSION IN BER SENSE WITH FINITE BLOCKLENGTH**

In this section, by combining the strong converse and cryptographic confusion and diffusion, a transmission scheme that is secure in the BER sense is proposed. Then the rate margins, security gains, and power optimization are discussed.

A. **Combining Strong Converse and Cryptographic Confusion and Diffusion**

When \( C_e < R < C_b \), by increasing blocklength \( n \), it is possible to make Bob’s block error probability arbitrarily small and Eve’s block error probability arbitrarily large. Ensuring small block error probability at Bob means reliable communication. However, ensuring high block error probability at Eve does not necessarily mean that the transmission is secure, because a
block error event simply means that there is at least a single bit error in the block. As a simple
example, one may consider the case where only a single bit within a codeword is always in error
whenever the codeword is decoded. In this case, the block error probability is one; however, all
other bits except the one are decoded by Eve, which means the communication is never secure.
In order to address this issue, a method to induce high BER at Eve is discussed.

The block diagram of the proposed scheme is presented in Fig. 3. Using an SPN, Alice
encrypts message $M$ of length $K$ bits into bit sequence $S^K$ of the same length, which is then encoded into a codeword $X^n$ of length $n$ symbols. Bob performs the inverse processing: he decodes the received codeword $Y^n_b$ into bit sequence $\hat{S}^K_b$, which is then decrypted into $\hat{M}_b$ by the inverse SPN. On the other hand, in principle, Eve can design her receiver as she wants, no matter what it is. In this paper, Eve’s receiver structure is assumed to be the same as Bob’s, which appears to be a reasonable assumption because otherwise it seems even more difficult for her to estimate $M$. Eve decodes the received codeword $Y^n_e$ into bit sequence $\hat{S}^K_e$, which is then decrypted into $\hat{M}_e$ by the inverse SPN.

At the receiver side (Bob or Eve), if no block decoding error occurs at the channel decoder, there is no bit error at the input of the inverse SPN, and thus, no output bit errors. On the other hand, when block decoding error occurs at the channel decoder, there is at least one bit error at the input of the inverse SPN and the input error(s) will be amplified by the inverse SPN. The BER $P_{\text{BER}}(R|C)$ for a code $C$ at the output of the inverse SPN is given by

$$P_{\text{BER}}(R|C) = P_{\text{BER}}^{\text{SPN}}(r, K) \big|_{\text{block error}} \times P_{\text{err}}(R|C)$$

where $P_{\text{err}}(R|C)$ denotes the block error probability at the output of the decoder and $P_{\text{BER}}^{\text{SPN}}(r, K) \big|_{\text{block error}}$ is the BER at the output of the inverse SPN given a block error. In order to (analytically or numerically) compute the BER $P_{\text{BER}}^{\text{SPN}}(r, K) \big|_{\text{block error}}$, the initial error distribution $q_{W_0}(w_0)$ must be determined from the condition that there was a block error, which means that there was at least a single bit error at the input of the inverse SPN. However, the exact number of bit errors within a block is random and the exact distribution of the number of bit errors is unknown. Furthermore, the exact block error probabilities, $P_{\text{err}}(R|C)$, for Bob and Eve are unknown. In the following, therefore, we consider their bounds: for Bob, an upper-bound of the ensemble average $\mathbb{E}[P_{\text{err}}(R|C)]$ is used; and for Eve, a lower-bound of $P_{\text{err}}(R|C)$ is used.

For Bob, using ensemble average block error probability upper-bound $P_{\text{err}}^b(R) \leq P_{\text{err}}^{b,U}(R, \hat{\rho}, \hat{q}(x))$ in (2) and noting that $P_{\text{BER}}^{\text{SPN}}(r, K) \big|_{\text{block error}}$ is upper-bounded by $0.5^4$ the ensemble average BER of Bob is upper-bounded as follows:

$$P_{\text{BER}}^b(R) \leq 0.5 P_{\text{err}}^{b,U}(R, \hat{\rho}, \hat{q}(x)) =: P_{\text{BER}}^{b,U}(R, \hat{\rho}, \hat{q}(x)).$$

Although $0.5$ is a trivial BER upper-bound, it is actually tight in our case because the output BER $P_{\text{BER}}^{\text{SPN}}(r, K) \big|_{\text{block error}}$ of SPN given a block error is (very) close to $0.5$ as long as $r$ is large enough, e.g., $r \geq 10$, as shown in Figs. 1 and 2.
For Eve, using $P_{\text{err}}(R|C) \geq P_{\text{err}}(R, \hat{\rho}', \hat{q}'(x)), \forall C$ in (14) and $P_{\text{BER}}^{\text{SPN}}(r, K) \geq P_{\text{BER}}^{\text{SPN,L}}(r, K)$, the BER is lower-bounded as follows:

$$P_{\text{BER}}^{\text{L}}(R|C) \geq P_{\text{BER}}^{\text{SPN,L}}(r, K) P_{\text{err}}(R, \hat{\rho}', \hat{q}'(x)), \forall C$$

(30)

$$=: P_{\text{BER}}^{\text{e,L}}(R, \hat{\rho}', \hat{q}'(x))$$

(31)

where $P_{\text{BER}}^{\text{SPN,L}}(r, K)$ is given by

$$P_{\text{BER}}^{\text{SPN,L}}(r, K) = P_{\text{BER}}^{\text{SPN}}(r, K)|_{\text{only one input bit error}}.$$  

(32)

That is, $P_{\text{BER}}^{\text{SPN,L}}(r, K)$ denotes the BER at the output of the inverse SPN when there is only a single input bit error (rather than at least one input bit error), and $P_{\text{BER}}^{\text{SPN,L}}(r, K)$ can be obtained by analysis or simulation as in Section II.C.

In general, the optimal distribution $\hat{q}(x)$ making the upper bound tightest and the distribution $\hat{q}'(x)$ making the lower bound valid for any $C$ are not necessarily the same, i.e., $\hat{q}(x) \neq \hat{q}'(x)$. For symmetric DMCs, however, they are the same and given by equi-probable distributions.

**Lemma 1:** For symmetric DMCs including BSC, BEC, and binary input (BI)-AWGN, we have

$$\hat{q}(x) = \hat{q}'(x) = q_{\text{equ}}(x)$$

(33)

where $q_{\text{equ}}(x)$ is the equi-probable distribution.

**Proof:** See Appendix A.

In our scheme, two bounds are imposed at the same time given a single transmitter (Alice). Therefore, it is important to ensure the existence of such code satisfying both bounds. That is, it must be ensured that at least a code exists for which Bob’s BER (not Bob’s ensemble average BER) is upper-bounded by $P_{\text{BER}}^{b,U}(R, \bar{\rho}, \bar{q}(x))$ and Eve’s BER is lower-bounded by $P_{\text{BER}}^{e,L}(R, \bar{\rho}', \bar{q}'(x))$ at the same time. Such existence is shown in the following.

**Lemma 2:** When $C_e < R < C_b$ and a priori probabilities are the same,

$\exists C$ such that $P_{\text{BER}}^{b}(R|C) \leq P_{\text{BER}}^{b,U}(R, \bar{\rho}, \bar{q}(x))$ and $P_{\text{BER}}^{e}(R|C) \geq P_{\text{BER}}^{e,L}(R, \bar{\rho}', \bar{q}'(x)).$  

(34)

**Proof:** When $R < C_b$, there exists at least one code for which Bob’s BER is upper-bounded by $P_{\text{BER}}^{b,U}(R, \bar{\rho}, \bar{q}(x))$. Furthermore, when $R > C_e$ and a priori probabilities are the same, Eve’s BER for any code is lower-bounded by $P_{\text{BER}}^{e,L}(R, \bar{\rho}', \bar{q}'(x))$. Therefore, there must exist a code satisfying both.
For the asymptotic case of infinite blocklength, we have \( \lim_{n \to \infty} P_{\text{err}}^{b,U}(R, \hat{p}, \hat{q}(x)) = 0 \) and
\( \lim_{n \to \infty} P_{\text{err}}^{e,L}(R, \hat{p}', \hat{q}'(x)) = 1 \) when \( C_e < R < C_b \). Thus, Bob’s BER upper-bound and Eve’s BER lower-bound are asymptotically given by
\[
\lim_{n \to \infty} P_{\text{BER}}^{b,U}(R, \hat{p}, \hat{q}(x)) = 0, \quad R < C_b \tag{35}
\]
\[
\lim_{n \to \infty} P_{\text{BER}}^{e,L}(R, \hat{p}', \hat{q}'(x)) = P_{\text{BER}}^{\text{SPN,L}}(r, K), \quad R > C_e. \tag{36}
\]
Recall that \( P_{\text{BER}}^{\text{SPN,L}}(r, K) \) can be made very close to 0.5 by increasing the number \( r \) of rounds, as demonstrated in Figs. 1 and 2.

B. Rate Upper and Lower Bounds for Finite Blocklength

In practice, the blocklength \( n \) is finite, and thus, it is not possible to achieve \( P_{\text{BER}}^{b,U}(R, \hat{p}, \hat{q}(x)) \to 0 \) when \( R < C_b \). In this paper, therefore, Bob’s BER upper-bound is constrained to be smaller than a BER threshold, \( 0 < P_{\text{BER}}^{b,\text{Th}} \leq 0.5 \), as follows:
\[
P_{\text{BER}}^{b}(R) \leq P_{\text{BER}}^{b,U}(R, \hat{p}, \hat{q}(x)) \leq P_{\text{BER}}^{b,\text{Th}}. \tag{37}
\]
This condition will be referred to as the reliability condition. To adjust \( P_{\text{BER}}^{b,\text{Th}} \), it is possible to use a block error probability threshold \( 0 < P_{\text{err}}^{b,\text{Th}} \leq 1 \), which is related to \( P_{\text{BER}}^{b,\text{Th}} \) as follows:
\[
P_{\text{BER}}^{b,\text{Th}} = 0.5 P_{\text{err}}^{b,\text{Th}}. \tag{38}
\]
For high reliability, \( P_{\text{err}}^{b,\text{Th}} \) should be set small (e.g., \( 10^{-6} \)). Similar to Bob’s case, with finite blocklength \( n \), it is not possible to achieve \( P_{\text{BER}}^{e,L}(R, \hat{p}', \hat{q}'(x)) \to P_{\text{BER}}^{\text{SPN,L}}(r, K) \) for Eve when \( R > C_e \). Therefore, Eve’s BER lower-bound is constrained to be larger than a BER threshold, \( P_{\text{BER}}^{e,\text{Th}} \) with \( 0 < P_{\text{BER}}^{e,\text{Th}} < P_{\text{BER}}^{\text{SPN,L}}(r, K) \) as follows:
\[
P_{\text{BER}}^{e}(R|C) \geq P_{\text{BER}}^{e,L}(R, \hat{p}', \hat{q}'(x)) \geq P_{\text{BER}}^{e,\text{Th}}, \quad \forall C. \tag{39}
\]
This condition will be referred to as the security condition. To adjust \( P_{\text{BER}}^{e,\text{Th}} \), it is possible to use a block error probability threshold \( 0 < P_{\text{err}}^{e,\text{Th}} < 1 \), which is related to \( P_{\text{BER}}^{e,\text{Th}} \) as follows:
\[
P_{\text{BER}}^{e,\text{Th}} = P_{\text{BER}}^{\text{SPN,L}}(r, K) P_{\text{err}}^{e,\text{Th}}. \tag{40}
\]
For high security, \( P_{\text{err}}^{e,\text{Th}} \) should be set large (e.g., \( 0.999999 \)).

When the reliability condition is imposed, the highest possible rate is lower than \( C_b \). Also, when the security condition is imposed, the lowest possible rate is higher than \( C_e \). In the following, the rate differences are defined.
Definition 1: The rate margin from above is defined by
\[ \Delta R_b := C_b - R_{sup} \]
and the rate margin from below is defined by
\[ \Delta R_e := R_{inf} - C_e \], where the highest allowable transmission rate \( R_{sup} \) and the lowest allowable transmission rate \( R_{inf} \) are determined by
\[ R_{sup} = \sup_{0 \leq R < C_b} R \quad \text{subject to} \quad P_{BER}^{b,U}(R, \tilde{\rho}, \tilde{q}(x)) \leq P_{BER}^{b, Th} \]
\[ R_{inf} = \inf_{R > C_e} R \quad \text{subject to} \quad P_{BER}^{e,L}(R, \tilde{\rho}', \tilde{q}'(x)) \geq P_{BER}^{e, Th} . \]

In the following theorem, \( \Delta R_b \) and \( \Delta R_e \) are analyzed.

**Theorem 1:** For \( P_{err}^{b, Th} = 1 \), we have \( \Delta R_b = 0 \). For \( 0 < P_{err}^{b, Th} < 1 \), we have
\[ \Delta R_b = -\frac{1}{n\tilde{\rho}} \ln P_{err}^{b, Th} + C_b - \frac{1}{\tilde{\rho}} E_0^b(\tilde{\rho}, \tilde{q}(x)) \geq 0 \]
\[ \Delta R_e = \frac{1}{n\tilde{\rho}'} \ln (1 - P_{err}^{e, Th}) + \frac{1}{\tilde{\rho}'} E_0^e(\tilde{\rho}', \tilde{q}'(x)) - C_e \geq 0 \]
where optimal \( \tilde{\rho} \) is determined by
\[ \tilde{\rho} = \arg \max_{0 < \rho \leq 1} \{ E_0^b(\rho, \tilde{q}(x)) - \rho R_{sup} \} \].
For \( P_{err}^{e, Th} = 0 \), we have \( \Delta R_e = 0 \). For \( 0 < P_{err}^{e, Th} < 1 \), we have
\[ \Delta R_e = \frac{1}{n\tilde{\rho}'} \ln (1 - P_{err}^{e, Th}) + \frac{1}{\tilde{\rho}'} E_0^e(\tilde{\rho}', \tilde{q}'(x)) - C_e \geq 0 \]
where optimal \( \tilde{\rho}' \) is determined by
\[ \tilde{\rho}' = \arg \max_{-1 < \rho' < 0} \{ E_0^e(\rho', \tilde{q}'(x)) - \rho' R_{inf} \} \].
As \( n \to \infty \), both rate margins tend to zero: \( \Delta R_b \to 0 \) and \( \Delta R_e \to 0 \).

**Proof:** See Appendix B.

From Theorem 1, one can see that, when \( 0 < P_{err}^{b, Th} < 1 \), the rate margin from above \( \Delta R_b \) is always positive, inversely proportional to \( n \) and \( \tilde{\rho} \), and logarithmically inversely proportional to \( P_{err}^{b, Th} \). Thus, to reduce \( \Delta R_b \), it appears that increasing the blocklength would be more effective than increasing \( P_{err}^{b, Th} \). A similar observation can be made for the rate margin from below \( \Delta R_e \).

Let \( \Delta R = R_{sup} - R_{inf} \) denote the rate interval in which the actual transmit rate \( R \) can be chosen. When \( \Delta R > 0 \), it is possible for Alice to transmit data reliably and securely satisfying (37) and (39). However, if \( \Delta R < 0 \), it is not possible to choose a rate \( R \) satisfying both conditions at the same time, and the data transmission is suspended. Letting \( \Delta C = C_b - C_e \) denote the capacity interval, the difference between capacity and rate intervals is given by
\[ \Delta C - \Delta R = \Delta R_b + \Delta R_e > 0. \] For the case of fading channels, the intervals are random variables and we have \( \Pr(\Delta C < 0) < \Pr(\Delta R < 0) \), meaning that the data suspension probability increases with shorter blocklength and stronger reliability/security conditions.

**Remark 1:** Ideally, the rate margins should have been defined using the constraints \( P_{\text{BER}}^{b,\text{Th}}(R) \leq P_{\text{BER}}^{b,\text{U}}(R, \bar{\rho}, \bar{q}(x)) \leq P_{\text{BER}}^{b,\text{Th}}(R, \bar{\rho}, \bar{q}(x)) \geq P_{\text{BER}}^{e,\text{Th}}(R, \bar{\rho}', \bar{q}'(x)) \geq P_{\text{BER}}^{e,\text{L}}(R, \bar{\rho}', \bar{q}'(x)) \) as in Theorem 1. Thus, the results of Theorem 1 can be interpreted as follows: There exists at least one code whose rate margins from above and below are not larger than \( \Delta R_b \) and \( \Delta R_e \), respectively.

### C. Security Gap

For some specific codes over BI-AWGN channels, the security gap was defined as the difference between Bob’s received signal to noise ratio (SNR) required to ensure Bob’s BER smaller than a threshold and Eve’s received SNR required to ensure Eve’s BER larger than a threshold [19], [20]. In general, the smaller the security gap, the suitable and more efficient the code for secure communications based on the BER security measure. By simulating specifically designed punctured-LDPC codes for BI-AWGN channels, the authors of [19] numerically obtained the security gap for their own codes. Similarly, in [20], the security gap was numerically obtained by simulating some specific BCH and LDPC codes combined with scrambling/descrambling for BI-AWGN channels. In this subsection, a fundamental limit of the security gap for any code with finite blocklength is studied for our proposed secure communications of combining strong converse and error amplification.

Consider the unconstrained Gaussian channel, where the received signals at Bob and Eve are given by

\[ Y_{b,i} = X_i + \eta_{b,i}, \quad i = 1, \cdots, n \]  
\[ Y_{e,i} = X_i + \eta_{e,i}, \quad i = 1, \cdots, n \]  

where \( \eta_{b,i} \sim \mathcal{CN}(0, \sigma_b^2) \) and \( \eta_{e,i} \sim \mathcal{CN}(0, \sigma_e^2) \) represent AWGNs at Bob and Eve, respectively. The transmitted signal \( X_i \) is normalized such that \( \mathbb{E}[||X_i|^2] = 1 \). Then the SNRs at Bob and Eve, respectively, are given by \( \gamma_b = \frac{\mathbb{E}[|X_i|^2]}{\sigma_b^2} = \frac{1}{\sigma_b^2} \) and \( \gamma_e = \frac{\mathbb{E}[|X_i|^2]}{\sigma_e^2} = \frac{1}{\sigma_e^2} \). We now define the security gap as follows.

**Definition 2:** For AWGN channels, the security gap \( \Delta S \) is defined by

\[ \Delta S := 10 \log_{10} \frac{\gamma_b^{\inf}}{\gamma_e^{\sup}} \]  

\[ (51) \]
where the lowest SNR $\gamma_b^{\inf}$ for Bob and the highest SNR $\gamma_e^{\sup}$ for Eve are determined by

$$
\gamma_b^{\inf} = \inf_{\gamma_b > \gamma_0} \gamma_b \quad \text{subject to} \quad P_{\text{BER}}^{b,\text{U}}(R, \bar{\rho}, \bar{q}(x), \gamma_b) \leq P_{\text{BER}}^{b,\text{Th}}
$$

$$
\gamma_e^{\sup} = \sup_{0 \leq \gamma_e < \gamma_0} \gamma_e \quad \text{subject to} \quad P_{\text{BER}}^{e,\text{L}}(R, \bar{\rho}', \bar{q}'(x), \gamma_e) \geq P_{\text{BER}}^{e,\text{Th}}.
$$

In the above equations, $\gamma_0 = C_{\text{AWGN}}^{-1}(R)$, where $C_{\text{AWGN}}(\gamma) = \ln(1 + \gamma)$ denotes the capacity of AWGN channels.

In order to determine Bob’s tightest ensemble average BER upper-bound and Eve’s valid BER lower bound for any code, the optimal input distributions $\bar{q}(x)$ and $\bar{q}'(x)$ must be first determined by maximizing $E_0^b(\rho, q(x), \gamma_b)$ and minimizing $E_0^e(\rho', q'(x), \gamma_e)$, respectively. Such optimizations are generally challenging, because the optimizations should be numerically performed and the optimal distributions depend on $\gamma_b$, $\gamma_e$, and $R$ (through $\bar{\rho}$ and $\bar{\rho}'$). In this subsection, for analytical tractability, we choose the input distributions as $\mathcal{CN}(0, 1)$, which is denoted by $q_{\mathcal{CN}}(x)$. With $q_{\mathcal{CN}}(x)$, the upper-bound of Bob’s ensemble average BER and the lower-bound of Eve’s BER are given in closed-form as follows:

$$
P_{\text{BER}}^{b,\text{U}}(R, \rho, q_{\mathcal{CN}}(x), \gamma_b) = 0.5 \exp \left( -n \left\{ E_0^b(\rho, q_{\mathcal{CN}}(x), \gamma_b) - \rho R \right\} \right)
$$

$$
P_{\text{BER}}^{e,\text{L}}(R, \rho', q_{\mathcal{CN}}(x), \gamma_e) = P_{\text{BER}}^{\text{SPN},\text{L}}(r, K) \cdot (1 - \exp (-n \left\{ E_0^e(\rho', q_{\mathcal{CN}}(x), \gamma_e) - \rho' R \right\}))
$$

where

$$
E_0^b(\rho, q_{\mathcal{CN}}(x), \gamma_b) = -\ln \left( 1 + \frac{\gamma_b}{1 + \rho} \right)^{-\rho}, \quad 0 \leq \rho \leq 1
$$

$$
E_0^e(\rho', q_{\mathcal{CN}}(x), \gamma_e) = -\ln \left( 1 + \frac{\gamma_e}{1 + \rho'} \right)^{-\rho'}, \quad -1 < \rho' \leq 0.
$$

To determine $\Delta S$ for AWGN channels, the highest SNR $\gamma_b^{\inf}$ and the lowest SNR $\gamma_e^{\sup}$ are first obtained in the following lemma.

**Lemma 3:** The solutions to (52) and (53) with $\bar{q}(x) = \bar{q}'(x) = q_{\mathcal{CN}}(x)$ are given by

$$
\gamma_b^{\inf} = \begin{cases} 
\gamma_0, & \text{if } P_{\text{err}}^{b,\text{Th}} = 1 \\
g_b(\bar{\rho}), & \text{if } 0 < P_{\text{err}}^{b,\text{Th}} < 1
\end{cases}
$$

$$
\gamma_e^{\sup} = \begin{cases} 
\gamma_0, & \text{if } P_{\text{err}}^{e,\text{Th}} = 0 \\
g_e(\bar{\rho}'), & \text{if } 0 < P_{\text{err}}^{e,\text{Th}} < 1
\end{cases}
$$

where

$$
g_b(\rho) = (1 + \rho) \left( (P_{\text{err}}^{b,\text{Th}})^{-\frac{1}{\alpha_p}} e^R - 1 \right)
$$

$$
g_e(\rho') = (1 + \rho') \left( (1 - P_{\text{err}}^{e,\text{Th}})^{-\frac{1}{\alpha_p'}} e^R - 1 \right).$$
Optimal $\hat{\rho}$ and $\hat{\rho}'$ are determined by

$$
\hat{\rho} = \arg \min_{0 < \rho \leq 1} g_b(\rho) \quad (62)
$$

$$
\hat{\rho}' = \arg \max_{-1 < \rho' < 0} g_e(\rho'). \quad (63)
$$

The optimal solution $\hat{\rho}$ to (62) always exists for left-open interval $(0, 1]$ and $g_b(\hat{\rho}) > \gamma_0$. The optimal solution $\hat{\rho}'$ to (63) always exists for open interval $(-1, 0)$ and $g_b(\hat{\rho}') < \gamma_0$. Also $g_b(\hat{\rho}')$ is positive if and only if the following condition is satisfied:

$$
\left(1 - \frac{2}{n} \ln(1 - \mathcal{P}_{\text{err}}^{e, \text{Th}})\right) \left(1 - \mathcal{P}_{\text{err}}^{e, \text{Th}}\right)^{\frac{1}{n}} e^R > 1. \quad (64)
$$

Proof: See Appendix C.

From the lemma, we immediately have the following result.

**Theorem 2:** When $0 < \mathcal{P}_{\text{err}}^{b, \text{Th}} < 1$ and $0 < \mathcal{P}_{\text{err}}^{e, \text{Th}} < 1$, the security gap $\Delta S$ with $\bar{q}(x) = \bar{q}'(x) = q_{CN}(x)$ is given by

$$
\Delta S = 10 \log_{10} \frac{(1 + \hat{\rho}) \left( \left( \mathcal{P}_{\text{err}}^{b, \text{Th}} \right)^{-\frac{1}{n \hat{\rho}}} e^R - 1 \right)}{(1 + \hat{\rho}') \left( \left( \mathcal{P}_{\text{err}}^{e, \text{Th}} \right)^{-\frac{1}{n \hat{\rho}'}} e^R - 1 \right)} \quad (65)
$$

where $\hat{\rho}$ and $\hat{\rho}'$ are given by (62) and (63), respectively. □

It is not difficult to show $\lim_{n \to \infty} \Delta S = 0$, which one can expect. Also, if one takes a high SNR approximation assuming $\gamma_b \gg 1$ and $\gamma_e \gg 1$, it is easier to obtain analytical insights into the security gap. When $\gamma_b \gg 1$ and $\gamma_e \gg 1$, the upper-bound of Bob’s ensemble average BER and the lower-bound of Eve’s BER can be approximated as follows:

$$
P_{\text{BER}}^{b, \text{U}}(R, \rho, q_{CN}(x), \gamma_b) \simeq 0.5 \exp \left( -n \left\{ - \ln \left( \frac{\gamma_b}{1 + \rho} \right)^{-\rho} - \rho R \right\} \right), \quad 0 \leq \rho \leq 1 \quad (66)
$$

$$
P_{\text{BER}}^{e, \text{L}}(R, \rho', q_{CN}(x), \gamma_e) \simeq P_{\text{BER}}^{\text{SPN,L}}(r, K) \cdot \left( 1 - \exp \left( -n \left\{ - \ln \left( \frac{\gamma_e}{1 + \rho'} \right)^{-\rho'} - \rho' R \right\} \right) \right), \quad -1 < \rho' \leq 0. \quad (67)
$$

From the approximate BER bounds, the security gap is obtained as follows:

$$
\Delta S \simeq -\frac{1}{n \hat{\rho}} 10 \log_{10} \mathcal{P}_{\text{err}}^{b, \text{Th}} + \frac{1}{n \hat{\rho}'} 10 \log_{10} \left( 1 - \mathcal{P}_{\text{err}}^{e, \text{Th}} \right) + 10 \log_{10} \left( \frac{1 + \hat{\rho}}{1 + \hat{\rho}'} \right) \quad (68)
$$

where $0 < \hat{\rho} \leq 1$ and $-1 < \hat{\rho}' < 0$. From this expression, one can easily see that the security gap is inversely proportional to $n$ and logarithmically inversely proportional to $\mathcal{P}_{\text{err}}^{b, \text{Th}}$ and $(1 - \mathcal{P}_{\text{err}}^{e, \text{Th}})$. Note that it is incorrect to interpret (68) to mean that, because $R$ does not explicitly appear
\[ \Delta S \] becomes independent of \( R \) in high SNR. Since both \( \tilde{\rho} \) and \( \tilde{\rho}' \) depend on \( R \), the security gap \( \Delta S \) still depends on \( R \) in high SNR.

Remark 2 (Input distribution): Gaussian distribution does not necessarily maximize \( E_b^0(\rho, q(x), \gamma_b) \), \( 0 \leq \rho \leq 1 \) for all \( \gamma_b \) and \( R < C_b \). Thus, the ensemble average BER upper-bound \( 54 \) is not necessarily the tightest one. Nevertheless, the upper-bound is still valid in the sense that there exists a code for which Bob’s BER is upper-bounded by \( 54 \). Similarly, Gaussian distribution does not necessarily minimize \( E_e^0(\rho', q'(x), \gamma_e), -1 < \rho' \leq 0 \) for all \( \gamma_e \) and \( R > C_e \). In this case, the BER lower-bound might not be valid in the sense that the BER of some codes might not be lower-bounded by \( 55 \). Consequently, for the particular code(s) whose BER at Bob is upper-bounded by \( 54 \) for all \( \gamma_b \) and \( R < C_b \), the corresponding BER at Eve might not be always larger than \( 55 \) for all \( \gamma_e \) and \( R > C_e \). In this sense, Eve’s BER lower-bound of \( 55 \) is optimistic. Nevertheless, using Gaussian input distribution is still useful because it makes the analysis tractable and gives some insights. Furthermore, it satisfies the asymptotic property \( 18 \) of the distribution, which makes Eve’s block error probability lower-bound valid for all codes, as follows:

\[
\lim_{\rho' \uparrow 0} \frac{1}{\rho'} E_e^0(\rho', q_N(x), \gamma_e) = \ln (1 + \gamma_e) = C_e. \tag{69}
\]

This means that Gaussian input distribution makes \( 55 \) valid for any code when \( \rho' \uparrow 0 \), which is optimal \( \rho' \) when \( R \to C_e \) from above. Furthermore, Gaussian distribution satisfies the asymptotic property \( 11 \) of the distribution, which makes Bob’s ensemble average block error probability upper-bound tightest, as follows:

\[
\lim_{\rho \downarrow 0} \frac{1}{\rho} E_b^0(\rho, q_N(x), \gamma_b) = \ln (1 + \gamma_b) = C_b. \tag{70}
\]

This means that Gaussian input distribution makes \( 54 \) tightest when \( \rho \downarrow 0 \), which is optimal \( \rho \) when \( R \to C_b \) from below.

Remark 3 (M-ary input AWGN): For one-dimensional or two-dimensional \( M \)-ary discrete input AWGN channels with equi-input probabilities, the security gap \( \Delta S \) can be obtained by \( 51 \), \( 52 \), and \( 53 \) by using \( \gamma_0 = C_{MI-AWGN}^{-1}(R) \), where \( C_{MI-AWGN}(\gamma) \) denotes the capacity of the \( M \)-ary input AWGN channel given in \( 34 \) eq. (1.20)]. Also, Bob’s ensemble average BER upper-bound and Eve’s BER lower-bound can be obtained in a similar way as in the Gaussian input
case. As an example, for BI-AWGN, the bounds are given by

$$P_{\text{BER}}^{b,U}(R, \rho, q_{\text{equ}}(x), \gamma_b) = 0.5 \times \exp \left( -n \left\{-\ln \left( \int_{-\infty}^{\infty} \sqrt{\gamma_b} \frac{2\pi}{\exp\left(-\frac{1}{2}\gamma_b(y_b^2+1)\right) \left( \cosh\left(\frac{\gamma_b y_b}{1+\rho}\right) \right)^{1+\rho} dy_b} - \rho R \right) \right\} \right)$$

(71)

$$P_{\text{BER}}^{e,L}(R, \rho', q_{\text{equ}}(x), \gamma_e) = P_{\text{BER}}^{\text{SPN,L}}(r, K) \times \left(1 - \exp \left( -n \left\{-\ln \left( \int_{-\infty}^{\infty} \sqrt{\gamma_e} \frac{2\pi}{\exp\left(-\frac{1}{2}\gamma_e(y_e^2+1)\right) \left( \cosh\left(\frac{\gamma_e y_e}{1+\rho'}\right) \right)^{1+\rho'} dy_e} - \rho' R \right) \right\} \right) \right)$$

(72)

where $0 \leq \rho \leq 1$ and $-1 < \rho' \leq 0$ are optimized to obtain tightest bounds. Unlike the Gaussian input case, it is difficult to analytically obtain the security gap $\Delta S$ for the $M$-ary input case because the BER bounds are not given in closed-form. Thus, $\Delta S$ should be obtained numerically.

**Remark 4 (BSC and BEC):** Although the security gap was originally considered only for AWGN channels in the literature, the concept can be extended to other channels such as BSC and BEC. Let $\varepsilon_b$ denote the crossover and erasure probabilities for BSC and BEC, respectively, for Bob. Let $\varepsilon_e$ denote the crossover and erasure probabilities for BSC and BEC, respectively, for Eve. It is assumed that $0 \leq \varepsilon_b < \varepsilon_e \leq 0.5$ for BSC, and $0 \leq \varepsilon_b < \varepsilon_e \leq 1$ for BEC. Given $R$, the security gap can be defined as the difference between the two probabilities as follows:

**Definition 3:** For BSC and BEC, the security gap is defined as follows:

$$\Delta S := \varepsilon_{e}^{\inf} - \varepsilon_{b}^{\sup} \geq 0$$

(73)

where $\varepsilon_{b}^{\sup}$ and $\varepsilon_{e}^{\inf}$ are determined by

$$\varepsilon_{b}^{\sup} = \sup_{0 \leq \varepsilon_{b} < \varepsilon_{0}} \varepsilon_{b} \quad \text{subject to} \quad P_{\text{BER}}^{b,U}(R, \bar{\rho}, q_{\text{equ}}(x), \varepsilon_{b}) \leq P_{\text{BER}}^{b,\text{Th}}$$

(74)

$$\varepsilon_{e}^{\inf} = \inf_{\varepsilon_{e} > \varepsilon_{0}} \varepsilon_{e} \quad \text{subject to} \quad P_{\text{BER}}^{e,L}(R, \bar{\rho}', q_{\text{equ}}(x), \varepsilon_{e}) \geq P_{\text{BER}}^{e,\text{Th}}$$

(75)

In these equations, $\varepsilon_{0} = C_{\text{BSC}}^{-1}(R)$ for BSC and $\varepsilon_{0} = C_{\text{BEC}}^{-1}(R)$ for BEC, where $C_{\text{BSC}}(\varepsilon)$ and $C_{\text{BEC}}(\varepsilon)$ are the capacities of BSC and BEC, respectively.

□
For BSC, the BER bounds are given by

\[ P_{\text{BER}}^{b,U}(R, \rho, q_{\text{equ}}(x), \varepsilon_b) = 0.5 \exp \left( -n \left\{ -\ln \left( 2^{-\rho} \left( \varepsilon_b^{1+\rho} + (1 - \varepsilon_b)^{1+\rho} \right) \right) - \rho R \right\} \right) \]  

(76)

\[ P_{\text{BER}}^{e,L}(R, \rho', q_{\text{equ}}(x), \varepsilon_e) = P_{\text{BER}}^{\text{SPN,L}}(r, K) \cdot \left( 1 - \exp \left( -n \left\{ -\ln \left( 2^{-\rho'} \left( \varepsilon_e^{1+\rho'} + (1 - \varepsilon_e)^{1+\rho'} \right) \right) - \rho' R \right\} \right) \right) \]  

(77)

For BEC, the BER bounds are given by

\[ P_{\text{BER}}^{b,U}(R, \rho, q_{\text{equ}}(x), \varepsilon_b) = 0.5 \exp \left( -n \left\{ -\ln \left( 2^{-\rho} \left( 1 - \varepsilon_b \right) + \varepsilon_b \right) - \rho R \right\} \right) \]  

(78)

\[ P_{\text{BER}}^{e,L}(R, \rho', q_{\text{equ}}(x), \varepsilon_e) = P_{\text{BER}}^{\text{SPN,L}}(r, K) \cdot \left( 1 - \exp \left( -n \left\{ -\ln \left( 2^{-\rho'} \left( 1 - \varepsilon_e \right) + \varepsilon_e \right) - \rho' R \right\} \right) \right) \]  

(79)

**D. Power Optimization for Gaussian-Input Fading Channels**

In this subsection, the transmit power is optimized for Gaussian-input fading channels. Let \( h_b \) denote the channel from Alice to Bob and \( h_e \) the channel from Alice to Eve, where \( h_b \) and \( h_e \) are fixed over the duration of a codeword. The received signals at Bob and Eve are given by

\[ Y_{b,i} = h_b X_i + \eta_{b,i}, \quad i = 1, \ldots, n \]  

(80)

\[ Y_{e,i} = h_e X_i + \eta_{e,i}, \quad i = 1, \ldots, n \]  

(81)

where \( \eta_{b,i} \sim \mathcal{CN}(0, \sigma_b^2) \) and \( \eta_{e,i} \sim \mathcal{CN}(0, \sigma_e^2) \). The transmit power \( p \) is given by \( p = \mathbb{E}[|X_i|^2] \). Let \( \Gamma_b = \frac{|h_b|^2}{\sigma_b^2} \) denote Bob’s instantaneous channel SNR and \( \Gamma_e = \frac{|h_e|^2}{\sigma_e^2} \) denote Eve’s instantaneous channel SNR. When the input distribution is given by \( q_{\mathcal{CN}}(x) = \mathcal{CN}(0, p) \), the upper-bound of Bob’s ensemble average BER and the lower-bound of Eve’s BER are given in closed-form as follows:

\[ P_{\text{BER}}^{b,U}(R, \rho, q_{\mathcal{CN}}(x), \Gamma_b, p) = 0.5 \exp \left( -n \left\{ -\ln \left( 1 + \frac{p \Gamma_b}{1 + \rho} \right)^{-\rho} - \rho R \right\} \right), \quad 0 \leq \rho \leq 1 \]  

(82)

\[ P_{\text{BER}}^{e,L}(R, \rho', q_{\mathcal{CN}}(x), \Gamma_e, p) = P_{\text{BER}}^{\text{SPN,L}}(r, K) \cdot \left( 1 - \exp \left( -n \left\{ -\ln \left( 1 + \frac{p \Gamma_e}{1 + \rho'} \right)^{-\rho'} - \rho' R \right\} \right) \right), \quad -1 < \rho' \leq 0 \]  

(83)

Using these bounds, we first define the reliability, security, and overall outage probabilities as follows:
Definition 4: The reliability outage is declared whenever \( P_{\text{BER}}^{\text{b,U}}(R, \tilde{p}, q_{CN}(x), \Gamma_b, p) > P_{\text{BER}}^{\text{b,Th}} \), the security outage is declared whenever \( P_{\text{BER}}^{\text{e,L}}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, p) < P_{\text{BER}}^{\text{e,Th}} \), and the overall outage is declared whenever \( P_{\text{BER}}^{\text{b,U}}(R, \tilde{p}, q_{CN}(x), \Gamma_b, p) > P_{\text{BER}}^{\text{b,Th}} \) or \( P_{\text{BER}}^{\text{e,L}}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, p) < P_{\text{BER}}^{\text{e,Th}} \). The reliability, security, and overall outage probabilities are given by

\[
P_{\text{out}}^{\text{rel}}(R, p) = \Pr \left( P_{\text{BER}}^{\text{b,U}}(R, \tilde{p}, q_{CN}(x), \Gamma_b, p) > P_{\text{BER}}^{\text{b,Th}} \right)
\]

\[
P_{\text{out}}^{\text{sec}}(R, p) = \Pr \left( P_{\text{BER}}^{\text{e,L}}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, p) < P_{\text{BER}}^{\text{e,Th}} \right)
\]

\[
P_{\text{out}}^{\text{overall}}(R, p) = \Pr \left( P_{\text{BER}}^{\text{b,U}}(R, \tilde{p}, q_{CN}(x), \Gamma_b, p) > P_{\text{BER}}^{\text{b,Th}} \text{ or } P_{\text{BER}}^{\text{e,L}}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, p) < P_{\text{BER}}^{\text{e,Th}} \right).
\]

Now, the transmit power is optimized to minimize the reliability outage probability subject to an average power constraint and the security condition for Eve:

\[
\min_{p(\Gamma_b, \Gamma_e)} P_{\text{out}}^{\text{rel}}(R, p(\Gamma_b, \Gamma_e))
\]

subject to \( p(\Gamma_b, \Gamma_e) \geq 0 \)

\[
\mathbb{E}[p(\Gamma_b, \Gamma_e)] \leq p_{\text{av}}
\]

\[
P_{\text{BER}}^{\text{e,L}}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, p(\Gamma_b, \Gamma_e)) \geq P_{\text{BER}}^{\text{e,Th}}
\]

where the transmit power \( p(\Gamma_b, \Gamma_e) \) is denoted as an explicit function of \( \Gamma_b \) and \( \Gamma_e \).

When \( P_{\text{BER}}^{\text{b,Th}} = 0.5 \) or \( P_{\text{err}}^{\text{Th}} = 1 \), the reliability outage probability is always zero. Also, when \( P_{\text{BER}}^{\text{e,Th}} = 0 \) or \( P_{\text{err}}^{\text{Th}} = 0 \), the security constraint is degenerate. Therefore, focusing on \( 0 < P_{\text{err}}^{\text{Th}} < 1 \) and \( 0 < P_{\text{err}}^{\text{Th}} < 1 \), the optimal solution is derived in the following.

Theorem 3: For \( 0 < P_{\text{err}}^{\text{Th}} < 1 \) and \( 0 < P_{\text{err}}^{\text{Th}} < 1 \), the optimal solution to (87) is given by

\[
p_{\text{opt}}(\Gamma_b, \Gamma_e) = \begin{cases} 
  p_{\text{min}}(\Gamma_b, \tilde{\rho}), & \text{if } p_{\text{min}}(\Gamma_b, \tilde{\rho}) \leq p_{\text{max}}(\Gamma_e, \tilde{\rho}') \text{ and } p_{\text{min}}(\Gamma_b, \tilde{\rho}) \leq z_{\text{opt}} \\
  0, & \text{if } p_{\text{min}}(\Gamma_b, \tilde{\rho}) > p_{\text{max}}(\Gamma_e, \tilde{\rho}') \text{ or } p_{\text{min}}(\Gamma_b, \tilde{\rho}) > z_{\text{opt}}
\end{cases}
\]

where

\[
p_{\text{min}}(\Gamma_b, \rho) = \Gamma_b^{-1} g_b(\rho)
\]

\[
p_{\text{max}}(\Gamma_e, \rho') = \Gamma_e^{-1} g_e(\rho')
\]

\[
z_{\text{opt}} = \max\{z : z \geq 0, \mathbb{E}[p_{\text{opt}}(\Gamma_b, \Gamma_e)] \leq p_{\text{av}}\}.
\]

In the above equations, \( g_b(\rho), g_e(\rho'), \tilde{\rho}', \) and \( \tilde{\rho} \) are respectively given by (60), (61), (62), and (63).
**Proof:** See the Appendix D.

The optimal power $p_{\text{opt}}(\Gamma_b, \Gamma_e)$ derived in Theorem 3 can be intuitively explained as follows. Firstly, in order to avoid any reliability outage at Bob, at least certain amount of transmit power should be used. Given $\Gamma_b$, power $p_{\text{min}}(\Gamma_b, \rho)$ is the minimum instantaneous power required to satisfy Bob’s reliability condition $P^{bU}_{\text{BER}}(R, \rho, q_{CN}(x), \Gamma_b, p) \leq P^{b,\text{Th}}_{\text{BER}}$. However, when we consider Eve, too much transmit power leads to weak security, because with higher power she can more easily decode the codeword. In order to enhance security for Eve and eventually to avoid any security outage at Eve, less transmit power should be used to ensure lower SNR at Eve. Given $\Gamma_e$, power $p_{\text{max}}(\Gamma_e, \rho')$ is the maximum instantaneous allowable power to satisfy the security condition $P^{eL}_{\text{BER}}(R, \rho', q_{CN}(x), \Gamma_e, p) \geq P^{e,\text{Th}}_{\text{BER}}$. Overall, any transmit power in the interval $[p_{\text{min}}(\Gamma_b, \rho), p_{\text{max}}(\Gamma_b, \rho')]$ satisfies both reliability and security conditions. With the average power constraint, however, the transmit power must be set to a minimum possible level by $p_{\text{opt}}(\Gamma_b, \Gamma_e) = p_{\text{min}}(\Gamma_e, \rho)$, to minimize the reliability outage probability by most efficiently utilizing the power on average.

Secondly, the case of $p_{\text{opt}}(\Gamma_b, \Gamma_e) = 0$ in Theorem 3 can be explained as follows. When $p_{\text{min}}(\Gamma_b, \rho)$ is greater than $p_{\text{max}}(\Gamma_b, \rho')$, it is not possible to satisfy both the reliability and security requirements at the same time, and thus, the data transmission must be suspended i.e., $p_{\text{opt}}(\Gamma_b, \Gamma_e) = 0$. Furthermore, due to the average power constraint, the transmission is suspended by setting $p_{\text{opt}}(\Gamma_b, \Gamma_e) = 0$ whenever the minimum power required is too large, i.e., $p_{\text{min}}(\Gamma_b, \rho) > z$. Let $P_{\text{sus}}(R, p(\Gamma_b, \Gamma_e))$ denote the data transmission suspension probability given by

$$P_{\text{sus}}(R, p(\Gamma_b, \Gamma_e)) = \Pr(p(\Gamma_b, \Gamma_e) = 0). \quad (92)$$

Then $z$ is maximized under the average power constraint in order to minimize the suspension probability, which is a necessary condition for reliability outage probability minimization because reliability outage occurs whenever $p(\Gamma_b, \Gamma_e) = 0$. The condition of $p_{\text{min}}(\Gamma_b, \rho) > z_{\text{opt}}$ can be rewritten as

$$\Gamma_b < \frac{g_b(\rho)}{z_{\text{opt}}}. \quad (93)$$

This means that, for efficient power consumption, the data transmission must be suspended when Bob’s instantaneous channel SNR $\Gamma_b$ is worse than a threshold.

Finally, the reason why optimal $\bar{\rho}$ and $\bar{\rho}'$ in Theorem 3 are obtained by (62) and (63), respectively, can be explained as follows. In order to minimize $P_{\text{sus}}(R, p(\Gamma_b, \Gamma_e))$ or $\Pr(p(\Gamma_b, \Gamma_e) = 0)$,
the value of $\rho$ must be optimized to minimize $p_{\min}(\Gamma_b, \rho)$, which is equivalent to the optimization in (62). Also, $\rho'$ must be optimized to maximize $p_{\max}(\Gamma_e, \rho')$, which is equivalent to the optimization in (63). The computational complexities required for these optimization are not high because each of $\tilde{\rho}$ and $\tilde{\rho}'$ can be individually obtained by one-dimensional searching.

From Theorem 3, we also have the following result.

**Corollary 1:** With the optimal power $p_{\text{opt}}(\Gamma_b, \Gamma_e)$ of Theorem 3, we have

$$P_{\text{rel}}^{\text{out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e)) = P_{\text{overall}}^{\text{out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e)) = P_{\text{sus}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e))$$

$$P_{\text{sec}}^{\text{out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e)) = 0.$$  \hfill (94)

**Proof:** First, when $p_{\text{opt}}(\Gamma_b, \Gamma_e) = 0$, we have $\Pr \left( P_{e, \text{BER}}^{e, L}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, 0) < P^{e, \text{Th}_{\text{BER}}}_{e, \text{BER}} \right) = 0$, because $P_{e, \text{BER}}^{e, L}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, 0) = 0.5$ with probability one. Second, when $p_{\text{opt}}(\Gamma_b, \Gamma_e) = p_{\min}(\Gamma_b, \tilde{\rho})$, we have $\Pr \left( P_{e, \text{BER}}^{e, L}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, p_{\min}(\Gamma_b, \tilde{\rho})) < P^{e, \text{Th}_{\text{BER}}}_{e, \text{BER}} \right) = 0$, because $P_{e, \text{BER}}^{e, L}(R, \tilde{\rho}', q_{CN}(x), \Gamma_e, p_{\min}(\Gamma_b, \tilde{\rho})) = P^{e, \text{Th}_{\text{BER}}}_{e, \text{BER}}$ with probability one as shown in Appendix D. It follows from the total probability theorem that $P_{\text{sec}}^{\text{out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e)) = 0$. Given this, it is straightforward to show $P_{\text{rel}}^{\text{out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e)) = P_{\text{overall}}^{\text{out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e)) = P_{\text{sus}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e)).$  \hfill \Box

### IV. Numerical Results

![Fig. 4. Rate differences $\Delta R_b$ and $\Delta R_e$ for BSC with $\epsilon_b = 0.01$ and $\epsilon_e = 0.3.$](image-url)
In this section, we present numerical results for the proposed secure communication method. First, we present the numerical results of rate margins $\Delta R_b$ and $\Delta R_e$ obtained in Theorem 1. Different reliability and security requirements are tested by considering $\mathcal{P}_{b, Th}^{b, err} \in \{1, 0.01, 0.0001, 0.000001\}$ and $\mathcal{P}_{e, Th}^{e, err} \in \{0, 0.99, 0.9999, 0.999999\}$. Recall that Bob’s BER upper-bound threshold is given by $\mathcal{P}_{b, BER}^{b, Th} = 0.5 \mathcal{P}_{b, err}^{b, Th}$ and Eve’s BER lower-bound threshold is given by $\mathcal{P}_{e, BER}^{e, Th} = \mathcal{P}_{e, L}^{SPN}(r, K) \mathcal{P}_{e, err}^{e, Th}$. Fig. 4 shows $\Delta R_b$ and $\Delta R_e$ for BSC with $\varepsilon_b = 0.01$ and $\varepsilon_e = 0.3$ for different blocklengths $10^2 \leq n \leq 10^6$. Also, Fig. 5 shows $\Delta R_b$ and $\Delta R_e$ for BI-AWGN with $\gamma_b = 6$ dB and $\gamma_e = -2$ dB. One can see that as the blocklength $n$ increases, $\Delta R_b$ and $\Delta R_e$ approach zero as expected in Theorem 1. With weaker reliability and security requirements (i.e., larger $\mathcal{P}_{b, err}^{b, Th}$ and smaller $\mathcal{P}_{e, err}^{e, Th}$), the rate margins decrease.

Second, we present the numerical results for the security gap $\Delta S$ obtained in Theorem 2. Fig. 6 shows the region where the condition of (64) is satisfied. It can be easily seen that the condition is satisfied for all practical cases, e.g., for all $R$ with $n > 10$. Fig. 7 shows the security gaps for BI-AWGN with $R = 0.5$ (nats/one-dimensional-channel use) and Gaussian-input (GI) AWGN with $R = 1$ (nats/two-dimensional-channel use). It can be seen that, for the same reliability and security conditions, Gaussian input gives smaller security gap than binary input.

We now consider fading channels, where $|h_b|$ and $|h_e|$ are modeled by Rayleigh random
variables with $E[|h_b|^2] = 2$ and $E[|h_e|^2] = 1$ and we set $\sigma_b^2 = \sigma_e^2 = \sigma^2$. Generating the channels $10^5$ times, we numerically obtain the reliability outage probability $P_{\text{out}}^{\text{rel}}(R, p)$ of (84), the security outage probability $P_{\text{out}}^{\text{sec}}(R, p)$ of (85), and the overall outage probability $P_{\text{out}}^{\text{overall}}(R, p)$ of (86). Note that when $|h_b| < |h_e|$ (i.e., $C_b < C_e$), it is never possible to avoid both reliability and security outages at the same time no matter which rate $R$ and transmit power $p$ are used. In the following, therefore, we evaluate the outage probabilities only when $|h_b| > |h_e|$. The reliability and security conditions are set to $P_{\text{err}}^{b,\text{Th}} = 0.0001$ and $P_{\text{err}}^{e,\text{Th}} = 0.9999$. The blocklength $n$ is set to $10^5$. We first consider the case of equal (or constant) transmit power, where the transmit power $p$ is set to $p_{\text{av}}$. Fig. 8 shows the outage probabilities with the equal transmit power for different rates $R \in \{0.5, 3.0, 5.5\}$ (nats/two-dimensional-channel use). Given rate $R$, as $p_{\text{av}}/\sigma^2$ increases, the reliability outage probability decreases, whereas the security outage probability increases as can be expected. Consequently, the overall outage probability always remains high (e.g., larger than say 0.4), because both reliability and security outage probabilities cannot be decreased at the same time.

For the proposed optimal power allocation of Theorem 3, Fig. 9 shows the outage probabilities with the same system parameters of Fig. 8. The obtained security outage probability of the optimal power allocation is exactly zero, which cannot be plotted in the figure of log-scale.
outage probability. The reliability probability, the overall outage probability, and the suspension probability are the same, as expected from Corollary 1. As $p_{av}/\sigma^2$ increases, the overall outage probability (or suspension probability) decreases and then flattens. However, the achieved lowest overall outage probability of the optimal power allocation is much lower than that of the constant transmit power case. The reason why there is an error floor for the overall outage probability or suspension probability is as follows. The suspension probability is given by $\Pr(p_{opt}(\Gamma_b, \Gamma_e) = 0)$. From Theorem 3, the suspension probability can be decreased only by reducing the probabilities of $p_{\min}(\Gamma_b, \tilde{\rho}) > p_{\max}(\Gamma_e, \tilde{\rho}')$ and $p_{\min}(\Gamma_b, \tilde{\rho}) > z_{opt}$. By increasing $p_{av}$, it is possible to reduce the probability of $p_{\min}(\Gamma_b, \tilde{\rho}) > z_{opt}$. However, it is not possible to reduce the probability of $p_{\min}(\Gamma_b, \tilde{\rho}) > p_{\max}(\Gamma_e, \tilde{\rho}')$ by increasing power $p_{av}$. In order to reduce the probability of $p_{\min}(\Gamma_b, \tilde{\rho}) > p_{\max}(\Gamma_e, \tilde{\rho}')$, Bob’s channel must be made even better (i.e., larger $\Gamma_b$) or Eve’s channel must be made even worse (i.e., smaller $\Gamma_e$). Only in this case, the suspension outage probability is further decreased. Fig. 10 shows the overall outage probabilities for the equal power and optimal power allocations for such channel scenario where Bob’s channel is much

---

\[ E_{\text{opt}} = \frac{1}{2} \Gamma_b \]
better than Eve’s channel: $E[|h_b|^2] = 10$ and $E[|h_e|^2] = 1$. In this channel scenario, the floors of the overall outage probabilities are lower for both equal and optimal power allocations. But, the performance gap between the two power allocations is still significant.

V. CONCLUSIONS

In this paper, a secure data transmission method has been studied, where the security measure was given in terms of the BER at the eavesdropper. To realize such secure communication, information-theoretic strong converse and cryptographic error amplification have been combined. For finite blocklengths, the maximum and minimum allowable transmission rates and the security gap have been analyzed for any block codes over DMCs. It has been observed that increasing the blocklength is very effective to reduce the rate loss and the security gap. For fading channels, the transmission power has been optimized. It has been found that simply increasing the transmission power does not decrease the reliability outage probability indefinitely. The error floor of the reliability outage probability depends on the channel quality difference between Bob and Eve.
Fig. 9. Reliability outage probability $P_{\text{rel, out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e))$ of (58), security outage probability $P_{\text{sec, out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e))$ of (85), and overall outage probability $P_{\text{overall, out}}(R, p_{\text{opt}}(\Gamma_b, \Gamma_e))$ of (86) with optimal power allocation in Theorem 3. $P_{\text{err, Th}}^b = 0.0001$ and $P_{\text{err, Th}}^e = 0.9999$. $E[|h_b|^2] = 2$ and $E[|h_e|^2] = 1$. Blocklength $n$ is $10^5$.

Fig. 10. Overall outage probabilities of equal transmit power and optimal transmit power. $P_{\text{err, Th}}^b = 0.0001$ and $P_{\text{err, Th}}^e = 0.9999$. $E[|h_b|^2] = 10$ and $E[|h_e|^2] = 1$. Blocklength $n$ is $10^5$. 

APPENDIX A
PROOF OF LEMMA 1

For symmetric DMCs, it is well-known that $E_0^0(\rho, q(x)), 0 \leq \rho \leq 1$, is maximized by equi-probable distribution $q_{equ}(x)$ [36 Theorem 7.2]. In the following, therefore, we will only show that $E_0^0(\rho', q'(x)), -1 < \rho' \leq 0$, is also minimized by $q_{equ}(x)$.

The proof of this appendix is only for finite input and output alphabet sizes. But, the approach holds for well-behaved channels with infinite alphabet sizes. For $x \in \{a_1, \cdots, a_Q\}$, let us define $\alpha(y_e, q)$ as follows

$$
\alpha(y_e, q) = \sum_x q(x) f_{Y|X}(y_e|x)^{1/(1+\rho')}, \quad -1 < \rho' \leq 0 \tag{A.1}
$$

and

$$
\sum_k q_k f_{Y|X}(y_e|x)^{1/(1+\rho')}, \quad -1 < \rho' \leq 0 \tag{A.2}
$$

where $q = (q_1, \cdots, q_Q) = (q(x = a_1), \cdots, q(x = a_Q))$. Because $\alpha(y_e, q)$ is linear in $q$ and the function $\alpha^{1+\rho'}$ is concave in $\alpha$, $\alpha(y, q)^{1+\rho'}$ must be concave in $q$. Letting $F(\rho', q) = \exp(-E_0^0(\rho', q)) = \sum_{y_e} \alpha(y_e, q)$, the function $F(\rho', q)$ is concave, because it is the sum of concave functions. Then $F(\rho', q)$ has a minimum for some $q^0$.

Following [23, Theorem 4.4.1], the necessary and sufficient conditions that $F(\rho', q)$ is minimized at $q^0$ are

$$
\left. \frac{\partial F(\rho', q)}{\partial q_k} \right|_{q=q^0} \leq \lambda, \quad k = 1, 2, \cdots, Q; -1 < \rho' \leq 0 \tag{A.3}
$$

where $\lambda$ is a constant and the equality holds whenever $q_k \neq 0$ (i.e., $q_k > 0$). Using [23 Theorem 5.6.5], the necessary and sufficient conditions on $q$ which maximize $F(\rho', q)$, equivalently, minimize $E_0^0(\rho', q)$, are

$$
\sum_{y_e} f_{Y|X}(y_e|x)^{1/(1+\rho')} \alpha(y_e, q)^{\rho'} \leq \sum_{y_e} \alpha(y_e, q)^{1+\rho'}, \quad -1 < \rho' \leq 0 \tag{A.4}
$$

where equality holds for which $q_k > 0$. Finally, for symmetric DMCs, the equi-probable distribution $q_k = \frac{1}{Q}$ satisfies the following condition [36 Theorem 7.2]:

$$
\sum_{y_e} f_{Y|X}(y_e|x)^{1/(1+\rho')} \alpha(y_e, q)^{\rho'} = \sum_{y_e} \alpha(y_e, q)^{1+\rho'}, \quad -1 < \rho' \leq 0. \tag{A.5}
$$

This complete the proof.
APPENDIX B
PROOF OF THEOREM 1

For $R_{sup}$, the optimization problem is

$$\sup_{0 \leq R < C_b} R \text{ subject to } \min_{0 \leq \rho \leq 1} P_{err}^{b,U}(R, \rho, \breve{q}(x)) \leq \mathcal{P}_{err}^{b,Th}(x)$$

where $P_{err}^{b,U}(R, \rho, \breve{q}(x)) = 2P_{BER}^{b,U}(R, \rho, \breve{q}(x))$. First, we consider the case of $\mathcal{P}_{err}^{b,Th} = 1$. In this case, the constraint is always satisfied. Thus, $R_{sup} = \sup_{0 \leq R < C_b} R = C_b$. Second, we consider the case of $0 < \mathcal{P}_{err}^{b,Th} < 1$. For $\rho = 0$, we have $P_{err}^{b,U}(R, \rho, \breve{q}(x)) = 1$. However, we know that $\min_{0 \leq \rho \leq 1} P_{err}^{b,U}(R, \rho, \breve{q}(x)) < 1$ because $\max_{0 \leq \rho \leq 1}\{E_0^b(\rho, \breve{q}(x)) - \rho R\} > 0$ for $R < C_b$. Thus, the optimal $\tilde{\rho}$ must be in $0 < \tilde{\rho} \leq 1$. That is, we have $\tilde{\rho}(R) = \arg \min_{0 < \rho \leq 1} P_{err}^{b,U}(R, \rho, \breve{q}(x)) = \arg \max_{0 \leq \rho \leq 1}\{E_0^b(\rho, \breve{q}(x)) - \rho R\}$, where $\{E_0^b(\rho, \breve{q}(x)) - \rho R\}$ is convex in $\rho \in (0, 1]$ for $R < C_b$ [23, Proof of Theorem 5.6.3]. Because $P_{err}^{b,U}(R, \rho, \breve{q}(x))$ is a monotonically decreasing function of $R$, the constraint must be satisfied with equality to maximize $R$: $P_{err}^{b,U}(R_{sup}, \rho, \breve{q}(x)) = \mathcal{P}_{err}^{b,Th}(x)$. Thus, we have

$$R_{sup} = \frac{1}{n\tilde{\rho}(R_{sup})} \ln \mathcal{P}_{err}^{b,Th}(x) + \frac{1}{\tilde{\rho}(R_{sup})} E_0^b(\tilde{\rho}(R_{sup}), \breve{q}(x))$$

$$= \frac{1}{n\tilde{\rho}(R_{sup})} \ln \mathcal{P}_{err}^{b,Th}(x) + \max_{q(x)} \frac{\partial}{\partial \rho} E_0^b(\rho, q(x)) \bigg|_{\rho = 0}$$

$$= \frac{1}{n\tilde{\rho}(R_{sup})} \ln \mathcal{P}_{err}^{b,Th}(x) + \max_{q(x)} I_b(q(x))$$

$$= \frac{1}{n\tilde{\rho}(R_{sup})} \ln \mathcal{P}_{err}^{b,Th}(x) + C_b$$

$$< C_b$$

where (b) is due to [15, eq.(34)] and (c) is valid for any $0 < \mathcal{P}_{err}^{b,Th} < 1$. When $n \to \infty$, we have $R_{sup} \to C_b$ from below, because the constraint becomes always satisfied by $\min_{0 \leq \rho \leq 1} P_{err}^{b,U}(R, \rho, \breve{q}(x)) \to 0$ for any $R < C_b$ as $n \to \infty$. Also, when $n \to \infty$, we have $\Delta R_b \to 0$.

For $R_{inf}$, the optimization problem is

$$\inf_{R > C_e} R \text{ subject to } \max_{-1 < \rho' \leq 0} P_{err}^{e,L}(R, \rho', \breve{q}(x)) \geq \mathcal{P}_{err}^{e,Th}(x)$$

where $P_{err}^{e,L}(R, \rho', \breve{q}(x)) = P_{BER}^{e,L}(R, \rho', \breve{q}(x))/\mathcal{P}_{BER}^{SPN,L}(r, K)$. We first consider the case of $\mathcal{P}_{err}^{e,Th} = 0$. In this case, the constraint is always satisfied. Thus, $R_{inf} = \inf_{R > C_e} R = C_e$. Second,
consider the case of \( 0 < P_{\text{err}}^{\text{e},\text{Th}} < 1 \). For \( \rho' = 0 \), we have \( P_{\text{err}}^{\text{e},U}(R, \rho', \bar{q}(x)) = 0 \). However, we know that \( \max_{-1 < \rho' \leq 0} P_{\text{err}}^{\text{e},U}(R, \rho', \bar{q}(x)) > 0 \) because \( \max_{-1 < \rho' \leq 0} \{ E_0(\rho', \bar{q}(x)) - \rho'R \} > 0 \) for \( R > C_e \). Thus, the optimal \( \rho' \) must be in \( -1 < \rho' < 0 \). That is, we have \( \bar{p}(R) = \arg \max_{-1 < \rho' < 0} P_{\text{err}}^{\text{e},U}(R, \rho', \bar{q}(x)) = \arg \max_{-1 < \rho' < 0} \{ E_0(\rho', \bar{q}(x)) - \rho'R \} \), where \( \{ E_0(\rho', \bar{q}(x)) - \rho'R \} \) is convex in \( \rho' \in (-1, 0) \) for \( R > C_e \) \[15\], \[24\] Lemma 3.2.1. Because \( P_{\text{err}}^{\text{e},U}(R, \rho', \bar{q}(x)) \) is a monotonically decreasing function of \( R \), the constraint must be satisfied with equality to minimize \( R \): \( \max_{-1 < \rho' \leq 0} P_{\text{err}}^{\text{e},U}(R_{\text{inf}}, \rho', \bar{q}(x)) = P_{\text{err}}^{\text{e},\text{Th}} \). Thus, we have

\[
R_{\text{inf}} = \frac{1}{n \bar{p}'(R_{\text{inf}})} \ln \left( 1 - P_{\text{err}}^{\text{e},\text{Th}} \right) + \frac{1}{\bar{p}'(R_{\text{inf}})} E_0(\bar{p}'(R_{\text{inf}}), \bar{q}(x)) \tag{B.9}
\]

\[
= \frac{1}{n \bar{p}'(R_{\text{inf}})} \ln \left( 1 - P_{\text{err}}^{\text{e},\text{Th}} \right) + \frac{1}{\bar{p}'(R_{\text{inf}})} \min E_0(\bar{p}'(R_{\text{inf}}), q(x)) \tag{B.10}
\]

\[
\overset{(d)}{=} \frac{1}{n \bar{p}'(R_{\text{inf}})} \ln \left( 1 - P_{\text{err}}^{\text{e},\text{Th}} \right) + \max_{q(x)} \frac{\partial}{\partial \rho} E_0(\bar{p}'(R_{\text{inf}}), q(x)) \bigg|_{\rho' = 0} \tag{B.11}
\]

\[
= \frac{1}{n \bar{p}'(R_{\text{inf}})} \ln \left( 1 - P_{\text{err}}^{\text{e},\text{Th}} \right) + \max_{q(x)} I_e(q(x)) \tag{B.12}
\]

\[
\overset{(e)}{=} C_e \tag{B.13}
\]

where \( (d) \) is due to \[15\] eq.(37) and \( (e) \) is valid for any \( 0 < P_{\text{err}}^{\text{e},\text{Th}} < 1 \). When \( n \to \infty \), we have \( R_{\text{inf}} \to C_e \) from above, because the constraint becomes always satisfied by \( \min_{-1 < \rho' \leq 0} P_{\text{err}}^{\text{e},U}(R, \rho', \bar{q}(x)) \to 1 \) for any \( R > C_e \) as \( n \to \infty \). Thus, when \( n \to \infty \), we have \( \Delta R_{\text{e}} \to 0 \).

**APPENDIX C**

**PROOF OF THEOREM 2**

For \( \gamma_b^{\text{inf}} \), the optimization problem is

\[
\inf_{\gamma_b > \gamma_0} \gamma_b \quad \text{subject to} \quad \min_{0 \leq \rho \leq 1} P_{\text{err}}^{\text{b},U}(R, \rho, q_{\text{CN}}(x), \gamma_b) \leq P_{\text{err}}^{\text{b},\text{Th}} \tag{C.1}
\]

where \( P_{\text{err}}^{\text{b},U}(R, \rho, q_{\text{CN}}(x), \gamma_b) = 2P_{\text{BER}}^{\text{b},U}(R, \rho, q_{\text{CN}}(x), \gamma_b) \). First, we consider the case of \( P_{\text{err}}^{\text{b},\text{Th}} = 1 \). In this case, the constraint is always satisfied. Thus, \( \gamma_b^{\text{inf}} = \inf_{\gamma_b > \gamma_0} \gamma_b = \gamma_0 \). Second, we consider the case of \( 0 < P_{\text{err}}^{\text{b},\text{Th}} < 1 \). As shown in Appendix B, the interval for optimizing \( \rho \) can be restricted to \( 0 < \rho \leq 1 \). For this interval, the constraint can be rewritten as

\[
\gamma_b \geq \min_{0 < \rho \leq 1} \left( 1 + \rho \right) \left( \frac{P_{\text{err}}^{\text{b},\text{Th}}}{\rho} e^{R} - 1 \right) = \min_{0 < \rho \leq 1} g_b(\rho). \tag{C.2}
\]
Thus, \( \gamma_b^\text{inf} = \inf_{\gamma_b \geq 0} \gamma_b = g_b(\hat{\rho}) \), where optimal \( \hat{\rho} \) is given by \( \hat{\rho} = \arg \min_{0 < \rho \leq 1} g_b(\rho) \). Finally, we show the existence of \( \hat{\rho} \) for left-open interval \((0, 1] \). For all \( 0 < \rho \leq 1 \), it is straightforward to show

\[
g_b(\rho) > \gamma_0 \geq 0
\]

\[
\lim_{\rho \to 0^+} \frac{\partial}{\partial \rho} g_b(\rho) = -\infty
\]

\[
\lim_{\rho \to 1^-} \frac{\partial}{\partial \rho} g_b(\rho) = u_b e^R - 1 > 0
\]

\[
\frac{\partial^2}{\partial \rho^2} g_b(\rho) > 0
\]

where \( u_b = \left( P_{\text{err}}^{b, \text{Th}} \right)^{-\frac{1}{2}} > 1 \). Therefore, the optimal \( \rho \) minimizing \( g_b(\rho) \) must not be \( \rho \to 0^+ \), and a solution exists in \((0, 1] \).

For \( \gamma_{e}^{\text{sup}} \), the optimization problem is

\[
\sup_{0 \leq \gamma_e < \gamma_0} \gamma_e \quad \text{subject to} \quad \max_{-1 < \rho' < 0} P_{\text{err}}^{e, L}(R, \rho', q_{\text{CN}}(x), \gamma_e) \geq P_{\text{err}}^{\text{Th}}
\]

where \( P_{\text{err}}^{e, L}(R, \rho', q_{\text{CN}}(x), \gamma_e) = P_{\text{BER}}^{e, L}(R, \rho', q_{\text{CN}}(x), \gamma_e)/P_{\text{BER}}^{\text{SN, L}}(r, K) \). First, we consider the case of \( P_{\text{err}}^{e, \text{Th}} = 0 \). In this case, the constraint is always satisfied. Thus, \( \gamma_{e}^{\text{sup}} = \sup_{0 \leq \gamma_e < \gamma_0} \gamma_e = \gamma_0 \). Second, we consider the case of \( 0 < P_{\text{err}}^{e, \text{Th}} < 1 \). As shown in Appendix B, the interval for optimizing \( \rho' \) can be restricted to \(-1 < \rho' < 0 \). For this interval, the constraint can be rewritten as

\[
\gamma_e \leq \max_{-1 < \rho' < 0} \left( (1 + \rho') \left( (1 - P_{\text{err}}^{e, \text{Th}})^{-\frac{1}{2}} e^R - 1 \right) \right) = \max_{-1 < \rho' < 0} \gamma_e(\rho').
\]

Thus, \( \gamma_{e}^{\text{sup}} = \sup_{0 \leq \gamma_e < \gamma_0} \gamma_e = g_e(\hat{\rho}') \), where optimal \( \hat{\rho}' \) is given by \( \hat{\rho}' = \arg \max_{-1 < \rho' < 0} \gamma_e(\rho') \).

Finally, we consider the existence of \( \hat{\rho}' \) for open interval \((-1, 0) \). It is straightforward to show

\[
g_e(\rho') < \gamma_0
\]

\[
\lim_{\rho' \to -1} g_e(\rho') = 0
\]

\[
\lim_{\rho' \to 0^-} g_e(\rho') = -1
\]

\[
\lim_{\rho' \to -1} \frac{\partial}{\partial \rho'} g_e(\rho') = (1 + 2 \ln u_e) u_e^{-1} e^R - 1
\]

where \( u_e = \left( 1 - P_{\text{err}}^{e, \text{Th}} \right)^{-\frac{1}{2}} > 1 \). From (C.10) and (C.11), \( g_e(\rho') \) cannot be maximized by \( \rho' \to 0^- \). If \( \lim_{\rho' \to -1} \frac{\partial}{\partial \rho'} g_e(\rho') > 0 \), from (C.10) and (C.11), it is clear that \( g_e(\rho') \) cannot be maximized by \( \rho' \to -1 \) and a solution must exist in \((-1, 0) \). Furthermore, in this case, the
maximum value $g_e(\rho')$ must be positive due to (C.10). On the other hand, if $\lim_{\rho' \to -1} \frac{\partial}{\partial \rho'} g_e(\rho) = ((1 + 2 \ln u_e)u_e^{-1}e^{R} - 1) \leq 0$, we have $g_e(\rho') = (1 + \rho') \left( u_e^{-1}e^{R} - 1 \right) < (1 + \rho') ((1 + 2 \ln u_e)u_e^{-1}e^{R} - 1) \leq 0$ for all $-1 < \rho' < 0$. Thus, it follows from (C.10) that a solution does not exist for the open interval of $(-1, 0)$.

**Appendix D**

**Proof of Theorem 3**

The optimization problem is

$$
\min_{p(\Gamma_b, \Gamma_e)} \text{Pr} \left( \min_{0 \leq \rho \leq 1} P^{b,U}_{\text{err}}(R, \rho, q_{CN}(x), \Gamma_b, p(\Gamma_b, \Gamma_e)) > P^{b,\text{Th}}_{\text{err}} \right) \tag{D.1a}
$$

subject to $p(\Gamma_b, \Gamma_e) \geq 0$

$$
\mathbb{E}[p(\Gamma_b, \Gamma_e)] \leq p_{\text{av}} \tag{D.1b}
$$

$$
\max_{-1 < \rho' \leq 0} P^{e,L}_{\text{err}}(R, \rho', q_{CN}(x), \Gamma_e, p(\Gamma_b, \Gamma_e)) \geq P^{e,\text{Th}}_{\text{err}} \tag{D.1c}
$$

where $P^{b,U}_{\text{err}}(R, \rho, q_{CN}(x), \Gamma_b, p(\Gamma_b, \Gamma_e)) = 2P^{b,U}_{\text{BER}}(R, \rho, q_{CN}(x), \Gamma_b, p(\Gamma_b, \Gamma_e))$ and $P^{e,L}_{\text{err}}(R, \rho', q_{CN}(x), \Gamma_e, p(\Gamma_b, \Gamma_e)) = P^{e,L}_{\text{BER}}(R, \rho', q_{CN}(x), \Gamma_e, p(\Gamma_b, \Gamma_e))/P^{\text{SPN,L}}_{\text{BER}}(r, K)$. As shown in Appendix B, the interval for optimizing $\rho$ can be restricted to $0 < \rho \leq 0$ and the interval for optimizing $\rho'$ can be restricted to $-1 < \rho' \leq 0$. In (D.1), $\rho'$ is optimized to minimize $P^{b,U}_{\text{err}}(R, \rho, q_{CN}(x), \Gamma_b, p)$.

But, this is equivalent to optimizing $\rho$ to minimize the outage probability. Also, in (D.1), $\rho'$ is optimized to maximize $P^{e,L}_{\text{err}}(R, \rho', q_{CN}(x), \Gamma_e, p)$, which maximizes the probability that the instantaneous security condition (D.1d) is satisfied. But, this is equivalent to optimizing $\rho'$ to minimize the outage probability, because an outage is declared whenever the condition is not satisfied. Therefore, the problem of (D.1) is equivalent to the following:

$$
\min_{0 < \rho \leq 1, -1 < \rho' < 0} \min_{p(\Gamma_b, \Gamma_e, \rho, \rho')} \text{Pr} \left( P^{b,U}_{\text{err}}(R, \rho, q_{CN}(x), \Gamma_b, p(\Gamma_b, \Gamma_e, \rho, \rho')) > P^{b,\text{Th}}_{\text{err}} \right) \tag{D.2a}
$$

subject to $p(\Gamma_b, \Gamma_e, \rho, \rho') \geq 0$

$$
\mathbb{E}[p(\Gamma_b, \Gamma_e, \rho, \rho')] \leq p_{\text{av}} \tag{D.2b}
$$

$$
P^{e,L}_{\text{err}}(R, \rho', q_{CN}(x), \Gamma_e, p(\Gamma_b, \Gamma_e, \rho, \rho')) \geq P^{e,\text{Th}}_{\text{err}} \tag{D.2d}
$$

where $p(\Gamma_b, \Gamma_e, \rho, \rho')$ is denoted as an explicit function of $\rho$ and $\rho'$.

First, we focus on the inner optimization over $p(\Gamma_b, \Gamma_e, \rho, \rho')$, and then we later solve the outer optimization over $\rho$ and $\rho'$. In order to solve the inner optimization problem, following the
Because the reliability outage occurs if and only if
\[ p_\text{min}(\Gamma_b, \rho, \rho') \geq 0 \]  
(D.3a)
subject to \( p_\text{min}(\Gamma_b, \rho, \rho') \geq 0 \)  
(D.3b)
\[ Pr_b^\text{U}(R, \rho, q_{CN}(x), \Gamma_b, p(\Gamma_b, \Gamma_e, \rho, \rho')) \leq Pr_\text{err}^b \]  
(D.3c)
\[ Pr_e^\text{L}(R, \rho', q_{CN}(x), \Gamma_e, p(\Gamma_b, \Gamma_e, \rho, \rho')) \geq Pr_\text{err}^e \]  
(D.3d)
It can be shown that, from the reliability condition of (D.3c), the solution to this optimization problem must be given in the form of \( p_{\text{min}}(\Gamma_b, \rho) = \frac{1}{\Gamma_b} g_b(\rho) \). Furthermore, from the security condition of (D.3d), \( p_{\text{min}}(\Gamma_b, \rho) \) can be a valid solution only when the following inequality is satisfied:
\[ p_{\text{min}}(\Gamma_b, \rho) \leq p_{\text{max}}(\Gamma_b, \rho') \]  
(D.4)
where \( p_{\text{max}}(\Gamma_e, \rho') = \frac{1}{\Gamma_e} g_e(\rho') \). Then, following [35, Proposition 4], the optimal solution to the inner optimization problem of (D.2) is given by
\[ p_{\text{opt}}(\Gamma_b, \Gamma_e, \rho, \rho') = \begin{cases} 
  p_{\text{min}}(\Gamma_b, \rho), & \text{if } p_{\text{min}}(\Gamma_b, \rho) \leq p_{\text{max}}(\Gamma_e, \rho') \text{ and } p_{\text{min}}(\Gamma_b, \rho) \leq z_{\text{opt}} \\
  0, & \text{if } p_{\text{min}}(\Gamma_b, \rho) \leq p_{\text{max}}(\Gamma_e, \rho') \text{ and } p_{\text{min}}(\Gamma_b, \rho) > z_{\text{opt}} \\
  0, & \text{if } p_{\text{min}}(\Gamma_b, \rho) > p_{\text{max}}(\Gamma_e, \rho') 
\end{cases} \]  
(D.5)
where \( z_{\text{opt}} \) is determined such that the average power constraint is satisfied:
\[ z_{\text{opt}} = \max\{z : z \geq 0, \mathbb{E}[p_{\text{opt}}(\Gamma_b, \Gamma_e, \rho, \rho')] \leq p_{\text{av}}\} \]  
(D.6)
We now solve the outer optimization of (D.2), i.e., optimizing \( \rho \) and \( \rho' \) to minimize the reliability outage probability:
\[ (\hat{\rho}, \hat{\rho}') = \min_{0 < \rho \leq 1, -1 < \rho' < 0} Pr \left( Pr_b^\text{U}(R, \rho, q_{CN}(x), \Gamma_b, p) > Pr_\text{err}^b \right) \bigg|_{p = p_{\text{opt}}(\Gamma_b, \Gamma_e, \rho, \rho')} \]  
(D.7)
Because the reliability outage occurs if and only if \( p_{\text{opt}}(\Gamma_b, \Gamma_e, \rho, \rho') = 0 \), we have
\[ (\hat{\rho}, \hat{\rho}') = \min_{0 < \rho \leq 1, -1 < \rho' < 0} Pr \left( p_{\text{min}}(\Gamma_b, \rho) > p_{\text{max}}(\Gamma_e, \rho') \text{ or } p_{\text{min}}(\Gamma_b, \rho) > z_{\text{opt}} \right) \]  
(D.8)
This joint optimization for \( \rho \) and \( \rho' \) is equivalent to two independent optimizations: \( \min_{0 < \rho \leq 1} p_{\text{min}}(\Gamma_b, \rho) \) and \( \max_{-1 < \rho' < 0} p_{\text{max}}(\Gamma_e, \rho') \), which are equivalent to (62) and (63), respectively. Note that these optimizations are independent of \( \Gamma_b \) and \( \Gamma_e \); that is, \( \hat{\rho} \) and \( \hat{\rho}' \) are independent of the instantaneous channels.
REFERENCES

[1] C. E. Shannon, “Communication theory of secrecy systems,” *Bell. Syst. Tech. J.*, vol. 28, no. 4, pp. 656–715, Oct. 1949.

[2] A. Wyner, “The wire-tap channel,” *Bell Syst. Tech. J.*, vol. 54, pp. 1355–1387, Oct. 1975.

[3] I. Csiszar and J. Kornet, “Broadcast channels with confidential messages,” *IEEE Trans. Inf. Theory*, vol. 24, pp. 339–348, May 1978.

[4] S. Leung-Yan-Cheong and M. Hellman, “The Gaussian wire-tap channel,” *IEEE Trans. Inf. Theory*, vol. 24, pp. 451–456, July 1978.

[5] A. Thangaraj, S. Dihidar, A.R. Calderbank, S. W. McLaughlin, and J.-M. Merolla, “Applications of LDPC codes to the wiretap channel,” *IEEE Trans. Inf. Theory*, vol. 53, pp. 2933–2945, Aug. 2007.

[6] V. Rathi, M. Andersson, R. Thobaben, J. Kliwer, and M. Skoglund, “Performance analysis and design of two edge-type LDPC does for the BEC wiretap channel,” *IEEE Trans. Inf. Theory*, vol. 59, pp. 1048–1064, Feb. 2013.

[7] H. Mahdavifar and A. Vardy, “Achieving the secrecy capacity of wiretap channels using polar codes,” *IEEE Trans. Inf. Theory*, vol. 57, pp. 6428–6443, Oct. 2011.

[8] A. Subramanian, A. Thangaraj, M. Bloch, and S. W. McLaughlin, “Strong secrecy on the binary erasure wiretap channel using arge-girth LDPC does,” *IEEE Trans. Inf. Forensics Sec.*, vol. 6, pp. 585–594, Sept. 2011.

[9] M. Cheraghchi, F. Didier, and A. Shokrollahi, “Invertible extractors and wiretap protocols,” *IEEE Trans. Inf. Theory*, vol. 58, pp. 1254–1274, Feb. 2012.

[10] M. Bloch and J. Barros, “Physical-layer security,” *Cambridge University Press*, 2011.

[11] U. M. Maurer, “The strong secret key rate of discrete random triples,” in *Communication and Cryptography–Two Sides of One Tapestry*, R. Blahut et al., Eds., Boston, MA, 1994, pp. 271–285.

[12] U. M. Maurer and S. Wolf, “Information-theoretic key agreement: From weak to strong secrecy for free,” in *Lecture Notes in Computer Science*. Berlin, Germany: Springer, 2000, vol. 1807, pp. 351–368.

[13] W.-C. Liao, T.-H. Chang, W.-K. Ma, and C.-Y. Chi, “QoS-based transmit beamforming in the presence of eavesdroppers: an optimized artificial-noise-aided approach,” *IEEE Trans. Signal Process.*, vol. 59, no. 3, pp. 1202–1216, Mar. 2011.

[14] J. Yang, I.-M. Kim, and D. I. Kim, “Optimal cooperative jamming multiuser broadcast channel with multiple eavesdroppers,” *IEEE Trans. Wireless Commun.*, vol. 12, pp. 2840–2852, June 2013.

[15] S. Arimoto, “On the converse to the coding theorem for discrete memoryless channels,” *IEEE Trans. Inf. Theory*, vol. 19, pp. 357–359, May 1973.

[16] R. Rajesh, S. M. Shah, and V. Sharma, “On secrecy above secrecy capacity,” in *Proc. IEEE International Conf. Communication System (ICCS)*, 2012, pp. 70–74.

[17] Y. Polyanskiy, H. V. Poor, and S. Verdu, “Channel coding rate in the finite blocklength regime,” *IEEE Trans. Inf. Theory*, vol. 56, pp. 2307–2359, May 2010.

[18] J.-C. Belfiore and F. Oggier, “Secrecy gain: a wiretap lattice code design,” in *Proc. International Symposium on Information Theory and its Applications (ISITA)*, 2010, pp. 174–178.

[19] D. Klink, J. Ha, S. W. McLaughlin, J. Barros, and B.-I. Kwak, “LPDC codes for the Gaussian wiretap channel,” *IEEE Trans. Inf. Forensics Sec.*, vol. 6, pp. 532–540, Sept. 2011.

[20] M. Baldi, M. Bianchi, and F. Chiara, “Coding with scrambling, concatenation, and HARQ for the AWGN wire-tap channel: a security gap analysis,” *IEEE Trans. Inf. Forensics Sec.*, vol. 7, pp. 883–894, June 2012.

[21] D. J. Torrieri, “The information-bit error rate for block codes,” *IEEE Trans. Commin.*, vol. 32, pp. 474–476, Apr. 1984.

[22] S. Leung-Yan-Cheong, “On a special class of wiretap channels,” *IEEE Trans. Inf. Theory*, vol. 23, pp. 625–627, Sept. 1997.
[23] R. G. Gallager, *Information Theory and Reliable Communications*. New York: Wiley, 1968.
[24] A. J. Viterbi and J. K. Omura, *Principles of Digital Communication and Coding*. New York: McGraw-Hill, 1979.
[25] D. R. Stinson, *Cryptography theory and practice*, 3rd Ed., Chapman & Hall/CRC, 2006.
[26] H. Feistel, “Cryptography and computer privacy,” *Scientific American*, vol. 228, pp. 15–23, 1973.
[27] A. F. Webster an S. E. Tavares, “On the design of S-boxes,” in *Proc. Advances in Cryptology: CRYPTO ’85*, Berlin, Springer-Verlag, 1986.
[28] W. Meier and O. Staffelbach, “Nonlinearity criteria for cryptographic functions,” in *Advances in Cryptography-Eurocrypt ’89 (Lecture Notes in Computer Science)*, Springer Verlag, 1990, pp. 549–562.
[29] B. Preneel, W. V. Leekwijck, L. V. Linden, R. Govaerts, and J. Vandewalle, “Propagation characteristics of boolean functions,” in *Advances in Cryptography-Eurocrypt 1990 (Lecture Notes in Computer Science)*, Springer Verlag, 1990, pp. 161–173.
[30] K. C. Cupta and P. Sarkar, “Construction of perfect nonlinear and maximally nonlinear multiple-output boolean functions satisfying higher order strict avalanche criteria,” *IEEE Trans. Inf. Theory*, vol. 50, pp. 2886–2893, Nov. 2004.
[31] T. W. Cusick and P. Stanica, *Cryptographic boolean functions and applications*, Academic Press, 2009.
[32] H. M. Hey and S. E. Tavares, “Avalanche characteristics of substitution-permutation encryption networks,” *IEEE Trans. Computers*, vol. 44, pp. 1131–1139, Sept. 1995.
[33] H. Shi, Y. Deng, and Y. Guan, “Analysis of the avalanche effect of the AES S box,” in *Proc. 2nd International Conference on Artificial Intelligence, Management Science and Electronic Commerce (AIMSEC)*, 2011.
[34] W. E. Ryan and S. Lim, *Channel Codes: classical and modern*, Cambridge University Press, 2009.
[35] G. Caire, G. Taricco, and E. Biglieri, Optimum power control over fading channels, *IEEE Trans. Inf. Theroy*, vol. 45, pp. 1468–1489, July 1999.
[36] F. Jelinek, *Probabilistic Information Theory*, New York: McGraw-Hill, 1968.