

# Error Exponents for Asymmetric Two-User Discrete Memoryless Source-Channel Coding Systems\*

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## Abstract

We study the transmission of two discrete memoryless correlated sources, consisting of a common and a private source, over a discrete memoryless multi-terminal channel with two transmitters and two receivers. At the transmitter side, the common source is observed by both encoders but the private source can only be accessed by one encoder. At the receiver side, both decoders need to reconstruct the common source, but only one decoder needs to reconstruct the private source. We hence refer to this system by the asymmetric 2-user source-channel coding system. We derive a universally achievable lossless joint source-channel coding (JSCC) error exponent pair for the 2-user system by using a technique which generalizes Csiszár's type-packing lemma (1980) for the point-to-point (single-user) discrete memoryless source-channel system. We next investigate the largest convergence rate of asymptotic exponential decay of the system (overall) probability of erroneous transmission, i.e., the system JSCC error exponent. We obtain lower and upper bounds for the exponent. As a consequence, we establish a JSCC theorem with single letter characterization and we show that the separation principle holds for the asymmetric 2-user scenario. By introducing common randomization, we also provide a formula for the tandem (separate) source-channel coding error exponent. Numerical examples show that for a large class of systems consisting of two correlated sources and an asymmetric multiple-access channel with additive noise, the JSCC error exponent considerably outperforms the corresponding tandem coding error exponent.

*Index Terms:* asymmetric 2-user source-channel system, broadcast channel, common and private message, common randomization, discrete memoryless correlated sources, error exponent, multiple-access channel, lossless joint source-channel coding, separation principle, tandem coding, type packing lemma.

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# 1 Introduction

Recently, the study of the error exponent (reliability function) for point-to-point (single-user) source-channel coding systems (with or without memory) has illustrated substantial superiority of joint source-channel coding (JSCC) over the traditional tandem coding (i.e., separate source and channel coding) approach (e.g., [8], [29], [30]). It is of natural interest to study the JSCC error exponent for multi-terminal source-channel systems.

In this work, we address the asymmetric 2-user source-channel coding system depicted in Fig. 1. Two discrete memoryless correlated source messages  $(\mathbf{s}, \mathbf{l}) \in \mathcal{S}^{\tau n} \times \mathcal{L}^{\tau n}$  drawn from a joint distribution  $Q_{SL} : \mathcal{S} \times \mathcal{L}$ , consisting of a common source messages  $\mathbf{s}$  and a private source message  $\mathbf{l}$  of length  $\tau n$ , are transmitted over a discrete memoryless asymmetric communication channel described by  $W_{YZ|UX} : \mathcal{U} \times \mathcal{X} \rightarrow \mathcal{Y} \times \mathcal{Z}$  with block codes of length  $n$ , where  $\tau > 0$  (measured in source symbol/channel use) is the overall transmission rate. The common source can be accessed by both encoders, but the private source can only be observed by one encoder (say, Encoder 1). In this set-up, the goal is to send the common information to both receivers, and send the private information to only one receiver (say, Decoder 1).

This asymmetric 2-user system can be used to model [23] interference channels with cognitive radio, an emerging and promising wireless technology where wireless systems, equipped with flexible software, dynamically adapt to their environment (by for example adjusting the modulation format or the coding scheme) to harness unemployed spectral capabilities [25, 26, 27, 12, 13]. For example, it can model the practical situation where audio and video signals are modulated and transmitted to two receivers over a cognitive interference channel (without secrecy constraints) [23], with the cognitive receiver needing to decode both audio and video signals while the non-cognitive receiver needing to only reconstruct the audio signal. Furthermore, it is worthy to point out that the asymmetric 2-user system is a generalization of the following two classical asymmetric multi-terminal scenarios which have been extensively studied in the literature.

- i.) The CS-AMAC system: If we remove Decoder 2 from Fig. 1, and let  $|\mathcal{Z}| = 1$ , then the channel reduces to a multiple-access channel  $W_{Y|UX}$ , and the coding problem reduces to transmitting two correlated sources (CS) over an asymmetric multiple-access channel (AMAC) with one receiver.
- ii.) The CS-ABC system: If we remove Encoder 2 from Fig. 1, and let  $|\mathcal{U}| = 1$ , then the channel reduces to a broadcast channel  $W_{YZ|X}$ , and the coding problem reduces to transmitting two CS over an asymmetric broadcast channel (ABC) with one transmitter.

The sufficient and necessary condition for the reliable transmission of CS over the AMAC – i.e., the lossless JSCC theorem for the CS-AMAC system – has been derived with single letter characterization in [4]. The capacity region of the ABC has been determined in [21], and the JSCC theorem for CS-ABC system with arbitrary transmission rate can also be analogously carried out (e.g., [17]). In this work, we

study a refined version of the JSCC theorem for the general asymmetric 2-user system (depicted in Fig. 1), by investigating the achievable JSCC error exponent pair (for two receivers) as well as the system JSCC error exponent, i.e., the largest convergence rate of asymptotic exponential decay of the system (overall) probability of erroneous transmission. We also apply our results to the CS-AMAC and CS-ABC systems.

We outline our results as follows. We first extend Csiszár’s type packing lemma [8] from a single-letter (1-dimension) type setting to a joint (2-dimensional) type setting. By employing the joint type packing lemma and generalized maximum mutual information (MMI) decoders, we establish achievable exponential upper bounds for the probabilities of erroneous transmission over an augmented 2-user channel  $W_{YZ|TUX}$  for a given triple of  $n$ -length sequences  $(\mathbf{t}, \mathbf{u}, \mathbf{x})$ ; see Theorem 1. Here, the augmented channel  $W_{YZ|TUX}$  is induced from the original 2-user channel  $W_{YZ|UX}$  by adding an auxiliary random variable (RV)  $T$  such that  $T$ ,  $(UX)$ , and  $(YZ)$ , form a Markov chain in this order. We introduce the RV  $T$  because we will employ superposition encoding which maps a source message pair  $(\mathbf{s}, \mathbf{l})$  to a codeword triplet  $(\mathbf{t}, \mathbf{u}, \mathbf{x})$ , where  $\mathbf{t}$  is the auxiliary superposition codeword. For the asymmetric 2-user system, since one of the encoders has full access to both sources, it knows the output of the other encoder. By properly designing the two (superposition) encoders, we apply Theorem 1 to establish a universally achievable error exponent pair for the two receivers (namely, the pair of exponents can be achieved by a sequence of source-channel codes independently of the statistics of the source and the channel); this generalizes Körner and Sgarro’s exponent pair for ABC coding (with uniformly distributed message sets) [22]. We also employ Theorem 1 to establish a lower bound for the system JSCC error exponent; see Theorem 2. Note that one consequence of our results is a sufficient condition (forward part) for the JSCC theorem. In addition, we use Fano’s inequality to prove a necessary condition (converse part) which coincides with the sufficient condition, and hence completes the JSCC theorem (Theorem 3). We next demonstrate that the separation principle holds for the 2-user system, i.e., there exists a separate source and channel coding system which can achieve optimality from the point of view of reliable transmissibility.

Using an approach analogous to [8], we also obtain an upper bound for the system JSCC error exponent (Theorem 4). As applications, we then specialize these results to the CS-AMAC and CS-ABC systems. The computation of the lower and upper bounds for the system JSCC error exponent is partially studied for the CS-AMAC system when the channel admits a symmetric conditional distribution.

We next study the tandem coding error exponent for the asymmetric 2-user system, which is the exponent resulting from separate and independent source and channel coding under common randomization. We derive a formula for the tandem coding error exponent in terms of the corresponding 2-user source error exponent and the asymmetric 2-user channel error exponent (Theorem 6). Finally, by numerically comparing the lower bound of the JSCC error exponent and the upper bound of the tandem coding error exponent, we illustrate that, as for the point-to-point systems ([29], [30]), JSCC can considerably outperform tandem coding in terms of error exponent for a large class of binary CS-AMAC systems with additive noise.

At this point we pause to mention some related works in the literature on the multi-terminal JSCC of CS. The JSCC theorem for transmitting two CS over a (symmetric) multiple access channel (where each encoder can only access one source) has been studied in [1, 7, 14, 19, 20, 28], and the JSCC theorem for transmitting two CS over a (symmetric) broadcast channel (where each decoder needs to reconstruct one source) has been addressed in [5, 17]. These works focus on the case when the overall transmission rate  $\tau$  is 1 and establish some sufficient and/or necessary conditions for which the sources can be reliably transmitted over the channel. However, for both (symmetric) systems, no matter whether the transmission rate  $\tau$  is 1 or not, a tight sufficient and necessary condition (JSCC theorem) with single-letter characterization is still unknown.

The rest of the paper is organized as follows. In Section 2, we introduce the notation and some basic facts regarding the method of types. A generalized joint type packing lemma is presented in Section 3. In Section 4, we establish a universally achievable error exponent pair for the 2-user system, as well as a lower and an upper bound for the system JSCC error exponent. A JSCC theorem with single-letter characterization is also given and we demonstrate that the reliable transmissibility condition can be achieved by separately performing source and channel coding. In Section 5, we apply our results to the CS-AMAC and CS-ABC systems. We partially address the computation of the bounds for the system JSCC error exponent in Section 6. In Section 7, we provide an expression for the tandem coding error exponent for the 2-user system and we then show that the JSCC error exponent can be strictly larger than the tandem coding error exponent for many CS-AMAC systems. Finally, we state our conclusions in Section 8.

## 2 Preliminaries

The following notation and conventions are adopted from [8, 10]. For any finite set (or alphabet)  $\mathcal{X}$ , the size of  $\mathcal{X}$  is denoted by  $|\mathcal{X}|$ . The set of all probability distributions on  $\mathcal{X}$  is denoted by  $\mathcal{P}(\mathcal{X})$ . The type of an  $n$ -length sequence  $\mathbf{x} \triangleq (x_1, x_2, \dots, x_n) \in \mathcal{X}^n$  is the empirical probability distribution  $P_{\mathbf{x}} \in \mathcal{P}(\mathcal{X})$  defined by

$$P_{\mathbf{x}}(a) \triangleq \frac{1}{n}N(a|\mathbf{x}), \quad a \in \mathcal{X},$$

where  $N(a|\mathbf{x})$  is the number of occurrences of  $a$  in  $\mathbf{x}$ . Let  $\mathcal{P}_n(\mathcal{X}) \subseteq \mathcal{P}(\mathcal{X})$  be the collection of all types of sequences in  $\mathcal{X}^n$ . For any  $P_X \in \mathcal{P}_n(\mathcal{X})$ , the set of all  $\mathbf{x} \in \mathcal{X}^n$  with type  $P_X$  is denoted by  $\mathbb{T}_{P_X}$ , or simply by  $\mathbb{T}_X$  if  $P_X$  is understood. We also call  $\mathbb{T}_{P_X}$  or  $\mathbb{T}_X$  a type class.

Similarly, the joint type of  $n$ -length sequences  $\mathbf{x} \in \mathcal{X}^n$  and  $\mathbf{y} \triangleq (y_1, y_2, \dots, y_n) \in \mathcal{Y}^n$  is the empirical joint probability distribution  $P_{\mathbf{xy}} \in \mathcal{P}(\mathcal{X} \times \mathcal{Y})$  defined by

$$P_{\mathbf{xy}}(a, b) \triangleq \frac{1}{n}N(a, b|\mathbf{x}, \mathbf{y}), \quad (a, b) \in \mathcal{X} \times \mathcal{Y}.$$

Let  $\mathcal{P}_n(\mathcal{X} \times \mathcal{Y}) \subseteq \mathcal{P}(\mathcal{X} \times \mathcal{Y})$  be the collection of all joint types of sequences in  $\mathcal{X}^n \times \mathcal{Y}^n$ . The set of all  $\mathbf{x} \in \mathcal{X}^n$  and  $\mathbf{y} \in \mathcal{Y}^n$  with joint type  $P_{XY} \in \mathcal{P}_n(\mathcal{X} \times \mathcal{Y})$  is denoted by  $\mathbb{T}_{P_{XY}}$ , or simply by  $\mathbb{T}_{XY}$ .

For any finite sets  $\mathcal{X}$  and  $\mathcal{Y}$ , the set of all conditional distributions  $V_{Y|X} : \mathcal{X} \rightarrow \mathcal{Y}$  is denoted by  $\mathcal{P}(\mathcal{Y}|\mathcal{X})$ . The conditional type of  $\mathbf{y} \in \mathcal{Y}^n$  given  $\mathbf{x} \in \mathbb{T}_{P_X}$  is the empirical conditional probability distribution  $P_{\mathbf{y}|\mathbf{x}} \in \mathcal{P}(\mathcal{Y}|\mathcal{X})$  defined by

$$P_{\mathbf{y}|\mathbf{x}}(b|a) = \frac{N(a, b|\mathbf{x}, \mathbf{y})}{N(a|\mathbf{x})},$$

whenever  $N(a|\mathbf{x}) > 0$ ; otherwise (if  $N(a|\mathbf{x}) = 0$ ) define  $P_{\mathbf{y}|\mathbf{x}}(b|a) = 0$ ,  $(a, b) \in \mathcal{X} \times \mathcal{Y}$ .

Let  $\mathcal{P}_n(\mathcal{Y}|P_X)$  be the collection of all conditional distributions  $V_{Y|X}$  which are conditional types of  $\mathbf{y} \in \mathcal{Y}^n$  given an  $\mathbf{x} \in \mathbb{T}_{P_X}$ . For any conditional type  $V_{Y|X} \in \mathcal{P}_n(\mathcal{Y}|P_X)$ , the set of all  $\mathbf{y} \in \mathcal{Y}^n$  for a given  $\mathbf{x} \in \mathbb{T}_{P_X}$  satisfying  $P_{\mathbf{y}|\mathbf{x}} = V_{Y|X}$  is denoted by  $\mathbb{T}_{V_{Y|X}}(\mathbf{x})$ , or simply by  $\mathbb{T}_{Y|X}(\mathbf{x})$ , which is also called a conditional type class ( $V$ -shell) with respect to  $\mathbf{x}$ .

For finite sets  $\mathcal{X}, \mathcal{Y}, \mathcal{Z}$  with joint distribution  $P_{XYZ} \in \mathcal{P}(\mathcal{X} \times \mathcal{Y} \times \mathcal{Z})$ , we use  $P_X, P_{XY}, P_{YZ|X}$ , etc, to denote the corresponding marginal and conditional probabilities induced by  $P_{XYZ}$ . Note that for a given joint type  $P_{XY} \in \mathcal{P}_n(\mathcal{X} \times \mathcal{Y})$ ,  $\mathbb{T}_{P_{Y|X}}(\mathbf{x}) = \{\mathbf{y} : (\mathbf{x}, \mathbf{y}) \in \mathbb{T}_{P_{XY}}\}$ . Note also that

$$\{P_X V_{Y|X} : P_X \in \mathcal{P}_n(\mathcal{X}), V_{Y|X} \in \mathcal{P}_n(\mathcal{Y}|P_X)\} = \mathcal{P}_n(\mathcal{X} \times \mathcal{Y}).$$

In addition, we denote

$$\mathcal{P}_n(\mathcal{Y}|\mathcal{X}) \triangleq \bigcup_{P_X \in \mathcal{P}_n(\mathcal{X})} \mathcal{P}_n(\mathcal{Y}|P_X) \subseteq \mathcal{P}(\mathcal{Y}|\mathcal{X}).$$

To distinguish between different distributions (or types) defined on the same alphabet, we use sub-subscripts, say,  $i, j$ , in  $P_{X_i}, P_{X_i Y_j}, \mathbb{T}_{X_i Y_j}$ , and so on. For example,  $\mathbb{T}_{X_i Y_j}$  is the type class of the joint type  $P_{X_i Y_j} \in \mathcal{P}_n(\mathcal{X} \times \mathcal{Y})$ . For any distribution  $P_{XYZ} \in \mathcal{P}(\mathcal{X} \times \mathcal{Y} \times \mathcal{Z})$ , we use  $H_{P_{XYZ}}(\cdot)$  and  $I_{P_{XYZ}}(\cdot; \cdot)$  to denote the entropy and mutual information under  $P_{XYZ}$ , respectively, or simply by  $H(\cdot)$  and  $I(\cdot; \cdot)$  if  $P_{XYZ}$  is understood.  $D(P_X \parallel Q_X)$  denotes the Kullback-Leibler divergence between distributions  $P_X, Q_X \in \mathcal{P}(\mathcal{X})$ .  $D(V_{Y|X} \parallel W_{Y|X}|P_X)$  denotes the Kullback-Leibler divergence between stochastic matrices (conditional distributions)  $V_{Y|X}, W_{Y|X} \in \mathcal{P}(\mathcal{Y}|\mathcal{X})$  conditional on distribution  $P_X \in \mathcal{P}(\mathcal{X})$ . For  $\mathbf{x} \in \mathcal{X}^n$ ,  $\mathbf{y} \in \mathcal{Y}^n$  and  $\mathbf{z} \in \mathcal{Z}^n$ , since the types  $P_{\mathbf{x}}, P_{\mathbf{x}\mathbf{y}}$  and  $P_{\mathbf{x}\mathbf{y}\mathbf{z}}$  can also be represented as distributions of dummy RV's, we define the empirical entropy and mutual information by  $H(\mathbf{x}) \triangleq H_{P_{\mathbf{x}}}(X)$ ,  $I(\mathbf{x}; \mathbf{y}) \triangleq I_{P_{\mathbf{x}\mathbf{y}}}(X; Y)$  and  $I(\mathbf{x}; \mathbf{y}|\mathbf{z}) \triangleq I_{P_{\mathbf{x}\mathbf{y}\mathbf{z}}}(X; Y|Z)$ . Given distributions  $P_X \in \mathcal{P}(\mathcal{X})$  and  $W_{Y|X} \in \mathcal{P}(\mathcal{Y}|\mathcal{X})$ , let  $P_X^{(n)}$  and  $W_{Y|X}^{(n)}$  be their  $n$ -dimensional product distributions. All logarithms and exponentials throughout this paper are in base 2. The following facts will be frequently used throughout this paper.

**Lemma 1** [10]

$$\text{i.) } |\mathcal{P}_n(\mathcal{X})| \leq (n+1)^{|\mathcal{X}|}, \quad |\mathcal{P}_n(\mathcal{Y}|\mathcal{X})| \leq (n+1)^{|\mathcal{Y}||\mathcal{X}|}.$$

ii.) For any  $P_X, Q_X \in \mathcal{P}_n(\mathcal{X})$ , we have

$$(n+1)^{-|\mathcal{X}|} 2^{nH_{P_X}(X)} \leq |\mathbb{T}_{P_X}| \leq 2^{nH_{P_X}(X)},$$

and

$$(n+1)^{-|\mathcal{X}|} 2^{-nD(P_X \| Q_X)} \leq Q_X^{(n)}(\mathbb{T}_{P_X}) \leq 2^{-nD(P_X \| Q_X)}.$$

iii.) For any  $\mathbf{x} \in \mathbb{T}_{P_X}$ ,  $\mathbf{y} \in \mathbb{T}_{V_{Y|X}}(\mathbf{x})$  and  $W_{Y|X}, V_{Y|X} \in \mathcal{P}_n(\mathcal{Y}|P_X)$ , we have

$$(n+1)^{-|\mathcal{X}||\mathcal{Y}|} 2^{nH_{P_X V_{Y|X}}(Y|X)} \leq |\mathbb{T}_{V_{Y|X}}(\mathbf{x})| \leq 2^{nH_{P_X V_{Y|X}}(Y|X)},$$

$$W_{Y|X}^{(n)}(\mathbf{y}|\mathbf{x}) = 2^{-n[D(V_{Y|X} \| W_{Y|X} | P_X) + H_{P_X V_{Y|X}}(Y|X)]},$$

and hence

$$(n+1)^{-|\mathcal{X}||\mathcal{Y}|} 2^{-nD(V_{Y|X} \| W_{Y|X} | P_X)} \leq W_{Y|X}^{(n)}(\mathbb{T}_{V_{Y|X}}(\mathbf{x})|\mathbf{x}) \leq 2^{-nD(V_{Y|X} \| W_{Y|X} | P_X)}.$$

### 3 A Joint Type Packing Lemma

Let us first recall Csiszár's type packing lemma for JSCC [8], which is an essential tool to establish an exponentially achievable upper bound for the JSCC probability of error over a discrete memoryless channel.

**Lemma 2** [8, Lemma 6] Given finite set  $\mathcal{A}$  and a sequence of positive integers  $\{m_n\}$ , for arbitrary (not necessarily distinct) types  $P_{A_i} \in \mathcal{P}_n(\mathcal{A})$ , and positive integers  $N_i$ ,  $i = 1, 2, \dots, m_n$  with

$$\frac{1}{n} \log_2 N_i < H_{P_{A_i}}(A) - \delta, \tag{1}$$

where

$$\delta \triangleq \frac{2}{n} [|\mathcal{A}|^2 \log_2(n+1) + \log_2 m_n + 1],$$

there exist  $m_n$  disjoint subsets

$$\Omega_i = \left\{ \mathbf{a}_p^{(i)} \right\}_{p=1}^{N_i} \subseteq \mathbb{T}_{A_i} \triangleq \mathbb{T}_{P_{A_i}}$$

such that

$$|\mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \Omega_k| \leq N_k 2^{-n[I_{P_{A_i} V_{A'|A}}(A; A') - \delta]}, \tag{2}$$

for every  $i, k, p$  and  $V_{A'|A} \in \mathcal{P}_n(\mathcal{A}|\mathcal{A})$ , with the exception of the case when both  $i = k$  and  $V_{A'|A}$  is the conditional distribution such that  $V_{A'|A}(a'|a)$  is 1 if  $a' = a$  and 0 otherwise.

Note that Lemma 2 is a generalization of the packing lemma in [10, p. 162, Lemma 5.1], where the later one is used for channel coding, while Lemma 2 is used for JSCC. Roughly and intuitively, if  $\mathbf{a}$  is a transmitted codeword, then the possible sequences decoded as  $\mathbf{a}$  can be seen as elements in the "sphere"

$\mathbb{T}_{V_{A'|A}}(\mathbf{a})$  “centered” at  $\mathbf{a}$  for some  $V_{A'|A}$ . Equation (2) in the packing lemma states that there exist disjoint sets  $\Omega_k$  with bounded cardinalities such that the size of the intersection between the sphere  $\mathbb{T}_{V_{A'|A}}(\mathbf{a})$  for every  $\mathbf{a} \in \Omega_i$  and every set  $\Omega_k$  is “exponentially small” compared with the size of each  $\Omega_k$ . So the packing lemma can be used to prove the existence of good codes that have an exponentially small probability of error.

We herein extend Csiszár’s above type packing lemma from the (1-dimensional) single-letter type setting to a (2-dimensional) joint type setting. This lemma will play a key role in establishing an exponentially achievable upper bound (in Theorem 1) for the probability of erroneous transmission for our asymmetric 2-user source-channel system.

**Lemma 3** (Joint Type Packing Lemma) Given finite sets  $\mathcal{A}$  and  $\mathcal{B}$ , a sequence of positive integers  $\{m_n\}$ , and a sequence of positive integers  $\{m'_{in}\}$  associated with every  $i = 1, 2, \dots, m_n$ , for arbitrary (not necessarily distinct) types  $P_{A_i} \in \mathcal{P}_n(\mathcal{A})$  and conditional types  $P_{B_j|A_i} \in \mathcal{P}_n(\mathcal{B}|P_{A_i})$ , and positive integers  $N_i$  and  $M_{ij}$ ,  $i = 1, 2, \dots, m_n$  and  $j = j(i) = 1, 2, \dots, m'_{in}$  with

$$\frac{1}{n} \log_2 N_i < H_{P_{A_i}}(A) - \delta, \quad (3)$$

and

$$\frac{1}{n} \log_2 M_{ij} < H_{P_{A_i}P_{B_j|A_i}}(B|A) - \delta, \quad (4)$$

where

$$\delta \triangleq \frac{2}{n} \left[ |\mathcal{A}|^2 |\mathcal{B}|^2 \log_2(n+1) + \log_2 m_n + \log_2(\max_i m'_{in}) + \log_2 12 \right],$$

there exist  $m_n$  disjoint subsets

$$\Omega_i = \left\{ \mathbf{a}_p^{(i)} \right\}_{p=1}^{N_i} \subseteq \mathbb{T}_{A_i}$$

such that

$$|\mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \Omega_k| \leq N_k 2^{-n} \left[ I_{P_{A_i}V_{A'|A}}(A;A') - \delta \right], \quad (5)$$

for every  $i, k, p$  and  $V_{A'|A} \in \mathcal{P}_n(\mathcal{A}|\mathcal{A})$ , with the exception of the case when both  $i = k$  and  $V_{A'|A}$  is the conditional distribution such that  $V_{A'|A}(a'|a)$  is 1 if  $a' = a$  and 0 otherwise; furthermore, for every  $\mathbf{a}_p^{(i)} \in \Omega_i$  and every  $i$ , there exist  $m'_{in}$  disjoint subsets

$$\Omega_{ij}(\mathbf{a}_p^{(i)}) = \left\{ (\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \right\}_{q=1}^{M_{ij}}$$

such that  $\mathbf{b}_{p,q}^{(j)} \in \mathbb{T}_{B_j|A_i}(\mathbf{a}_p^{(i)}) \triangleq \mathbb{T}_{P_{B_j|A_i}}(\mathbf{a}_p^{(i)})$  and

$$\left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \bigcup_{p'=1}^{N_k} \Omega_{kl}(\mathbf{a}_{p'}^{(k)}) \right| \leq N_k M_{kl} 2^{-n} \left[ I_{P_{A_i}B_jV_{A'B'|AB}}(A,B;A',B') - \delta \right], \quad (6)$$

$$\left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \bigcup_{p'=1}^{N_i} \Omega_{il}(\mathbf{a}_{p'}^{(i)}) \right| \leq M_{il} 2^{-n \left[ I_{P_{A_i B_j} V_{A'B'|AB}}(B; B'|A) - \delta \right]}, \quad (7)$$

for any  $i, j, k, l, p, q$  and  $V_{A'B'|AB} \in \mathcal{P}_n(\mathcal{A} \times \mathcal{B} | \mathcal{A} \times \mathcal{B})$ , with the exception of the case when both  $i = k$ ,  $j = l$  and  $V_{A'B'|AB}$  is the conditional distribution such that  $V_{A'B'|AB}(a', b' | a, b)$  is 1 if  $(a', b') = (a, b)$  and 0 otherwise.

The proof of the packing lemma is lengthy and is deferred to Appendix A. Compared with Lemma 2, it is seen that Csiszár's type packing lemma (Equation (5)) is incorporated in our extended packing lemma, and we emphasize that here we need (6) and (7) hold in addition to (5).

Similarly, for the 2-user channel, if  $(\mathbf{a}, \mathbf{b})$  is a pair of transmitted codewords, then the possible sequences decoded as  $(\mathbf{a}, \mathbf{b})$  can be seen as elements in the "sphere"  $\mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}, \mathbf{b})$  "centered" at  $(\mathbf{a}, \mathbf{b})$  for some  $V_{A'B'|AB}$ . As depicted in Fig. 2, Equation (6) (similarly to (7)) states that there exist disjoint sets  $\Omega_{kl} = \bigcup_{p'=1}^{N_k} \Omega_{kl}(\mathbf{a}_{p'}^{(k)})$  with bounded cardinalities such that the size of the intersection between the sphere  $\mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}, \mathbf{b})$  for every  $(\mathbf{a}, \mathbf{b}) \in \Omega_{ij}$  and every set  $\Omega_{kl}$  is "exponentially small" compared with the size of each  $\Omega_{kl}$ . Note also that the extended packing lemma is analogous to, but different from the one introduced by Körner and Sgarro in [22], which is used to prove a lower bound for the channel coding ABC exponent. Lemma 3 here is used for the asymmetric 2-user JSCC problem.

## 4 Transmitting CS over the Asymmetric 2-User Channel

### 4.1 System

Let  $\{W_{YZ|UX} : \mathcal{U} \times \mathcal{X} \rightarrow \mathcal{Y} \times \mathcal{Z}\}$  be a 2-user discrete memoryless channel with finite input alphabet  $\mathcal{U} \times \mathcal{X}$ , finite output alphabet  $\mathcal{Y} \times \mathcal{Z}$ , and a transition distribution  $W_{YZ|UX}(y, z | u, x)$  such that the  $n$ -tuple transition probability is

$$W_{YZ|UX}^{(n)}(\mathbf{y}, \mathbf{z} | \mathbf{u}, \mathbf{x}) = \prod_{i=1}^n W_{YZ|UX}(y_i, z_i | u_i, x_i),$$

where  $u \in \mathcal{U}$ ,  $x \in \mathcal{X}$ ,  $y \in \mathcal{Y}$ ,  $z \in \mathcal{Z}$ ,  $\mathbf{u} \triangleq (u_1, \dots, u_n) \in \mathcal{U}^n$ ,  $\mathbf{x} = (x_1, \dots, x_n) \in \mathcal{X}^n$ ,  $\mathbf{y} = (y_1, \dots, y_n) \in \mathcal{Y}^n$ , and  $\mathbf{z} \triangleq (z_1, \dots, z_n) \in \mathcal{Z}^n$ . Denote the marginal transition distributions of  $W_{YZ|UX}$  at its  $Y$ -output (respectively  $Z$ -output) by  $W_{Y|UX} \triangleq \sum_{\mathcal{Z}} W_{YZ|UX}$  (respectively  $W_{Z|UX} \triangleq \sum_{\mathcal{Y}} W_{YZ|UX}$ ). The marginal distributions of  $W_{YZ|UX}^{(n)}$  are denoted by  $W_{Y|UX}^{(n)}$  and  $W_{Z|UX}^{(n)}$ , respectively.

Consider two discrete memoryless CS with a generic joint distribution  $Q_{SL}(s, l)$  defined on the finite alphabet  $\mathcal{S} \times \mathcal{L}$  such that the  $k$ -tuple joint distribution is  $Q_{SL}^{(k)}(\mathbf{s}, \mathbf{l}) = \prod_{i=1}^k Q_{SL}(s_i, l_i)$ , where  $(s, l) \in \mathcal{S} \times \mathcal{L}$ , and  $(\mathbf{s}, \mathbf{l}) \triangleq ((s_1, l_1), \dots, (s_k, l_k)) \in \mathcal{S}^k \times \mathcal{L}^k$ . For each pair of source messages  $(\mathbf{s}, \mathbf{l})$  drawn from the above joint distribution, we need to transmit the *common message*  $\mathbf{s}$  over the channel  $W_{YZ|UX}$  to Receivers  $Y$  and  $Z$  and transmit the *private message*  $\mathbf{l}$  only to Receiver  $Y$ . A joint source-channel (JSC) code with block length  $n$  and positive transmission rate  $\tau$  (source symbol/channel use) for transmitting  $Q_{SL}$  through



$W_{YZ|UX}$  is a quadruple of mappings,  $(f_n, g_n, \varphi_n, \psi_n)$ , where  $f_n : \mathcal{S}^{\tau n} \times \mathcal{L}^{\tau n} \rightarrow \mathcal{X}^n$  and  $g_n : \mathcal{S}^{\tau n} \rightarrow \mathcal{U}^n$  are called encoders, and  $\varphi_n : \mathcal{Y}^n \rightarrow \mathcal{S}^{\tau n} \times \mathcal{L}^{\tau n}$  and  $\psi_n : \mathcal{Z}^n \rightarrow \mathcal{S}^{\tau n}$  are referred to as  $Y$ -decoder and  $Z$ -decoder, respectively; see Fig. 1.

The probabilities of  $Y$ - and  $Z$ -error are given by

$$P_{Y_e}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \triangleq \Pr(\{\varphi_n(Y^n) \neq (S^{\tau n}, L^{\tau n})\}) = \sum_{\mathcal{S}^{\tau n} \times \mathcal{L}^{\tau n}} Q_{SL}^{(\tau n)}(\mathbf{s}, \mathbf{l}) \sum_{\mathbf{y}: \varphi_n(\mathbf{y}) \neq (\mathbf{s}, \mathbf{l})} W_{Y|UX}^{(n)}(\mathbf{y}|\mathbf{u}, \mathbf{x}) \quad (8)$$

and

$$P_{Z_e}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \triangleq \Pr(\{\psi_n(Z^n) \neq S^{\tau n}\}) = \sum_{\mathcal{S}^{\tau n}} Q_S^{(\tau n)}(\mathbf{s}) \sum_{\mathbf{z}: \psi_n(\mathbf{z}) \neq \mathbf{s}} W_{Z|UX}^{(n)}(\mathbf{z}|\mathbf{u}, \mathbf{x}) \quad (9)$$

where  $\mathbf{x} \triangleq f_n(\mathbf{s}, \mathbf{l})$  and  $\mathbf{u} \triangleq g_n(\mathbf{s})$  are the corresponding codewords of the source message pair  $(\mathbf{s}, \mathbf{l})$  and the source message  $\mathbf{s}$ , and  $\mathbf{y}$  and  $\mathbf{z}$  are the received codewords at the Receivers  $Y$  and  $Z$ , respectively. Since we will study the exponential behavior of these probabilities using the method of types, it might be a better way to rewrite the probabilities of  $Y$ - and  $Z$ - error as a sum of probabilities of types

$$P_{i_e}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) = \sum_{P_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})} Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) P_{i_e}(\mathbb{T}_{SL}), \quad i = Y, Z, \quad (10)$$

where  $\mathbb{T}_{SL} \triangleq \mathbb{T}_{P_{SL}}$ , and

$$P_{Y_e}(\mathbb{T}_{SL}) = \frac{1}{|\mathbb{T}_{SL}|} \sum_{(\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{SL}} \sum_{\mathbf{y}: \varphi_n(\mathbf{y}) \neq (\mathbf{s}, \mathbf{l})} W_{Y|UX}^{(n)}(\mathbf{y}|\mathbf{u}, \mathbf{x}) \quad (11)$$

and

$$P_{Z_e}(\mathbb{T}_{SL}) = \frac{1}{|\mathbb{T}_{SL}|} \sum_{(\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{SL}} \sum_{\mathbf{z}: \psi_n(\mathbf{z}) \neq \mathbf{s}} W_{Z|UX}^{(n)}(\mathbf{z}|\mathbf{u}, \mathbf{x}). \quad (12)$$

We say that the JSCC error exponent pair  $(E_{AY}, E_{AZ})$  is achievable with respect to  $\tau > 0$  if there exists a sequence of JSC codes  $(f_n, g_n, \varphi_n, \psi_n)$  with transmission rate  $\tau$  such that the probabilities of  $Y$ -error and  $Z$ -error are simultaneously bounded by

$$P_{i_e}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \leq 2^{-n[E_{Ai} - \delta]}, \quad i = Y, Z \quad (13)$$

for  $n$  sufficiently large and any  $\delta > 0$ . As the point-to-point system, we denote the system (overall) probability of error by

$$P_e^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \triangleq \Pr\left(\{\varphi_n(Y^n) \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_n(Z^n) \neq S^{\tau n}\}\right), \quad (14)$$

where  $(S^{\tau n}, L^{\tau n})$  are drawn according to  $Q_{SL}^{(\tau n)}$ .

**Definition 1** Given  $Q_{SL}$ ,  $W_{YZ|UX}$  and  $\tau > 0$ , the system JSCC error exponent  $E_J(Q_{SL}, W_{YZ|UX}, \tau)$  is defined as supremum of the set of all numbers  $E$  for which there exists a sequence of JSC codes  $(f_n, g_n, \varphi_n, \psi_n)$  with blocklength  $n$  and transmission rate  $\tau$  such that

$$E \leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_e^{(n)}(Q_{SL}, W_{YZ|UX}, \tau). \quad (15)$$

Since the system probability of error must be larger than  $P_{Ye}^{(n)}$  and  $P_{Ze}^{(n)}$  defined by (8) and (9), and is also upper bounded by the sum of the two, it follows that for any sequence of JSC codes  $(f_n, g_n, \varphi_n, \psi_n)$

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_e^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) = \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \max(P_{Ye}^{(n)}, P_{Ze}^{(n)}). \quad (16)$$

## 4.2 Superposition Encoding for Asymmetric 2-User Channels

Given an asymmetric 2-user channel  $W_{YZ|UX}$ , at the encoder side, we can artificially augment the channel input alphabet by introducing an auxiliary (arbitrary and finite) alphabet  $\mathcal{T}$ , and then look at the channel as a discrete memoryless channel  $W_{YZ|TUX} = W_{YZ|UX}$  with marginal distributions  $W_{Y|TUX}$  and  $W_{Z|TUX}$  such that  $W_{YZ|TUX}(y, z|t, u, x) = W_{YZ|UX}(y, z|u, x)$  for any  $t \in \mathcal{T}$ ,  $u \in \mathcal{U}$ ,  $x \in \mathcal{X}$ ,  $y \in \mathcal{Y}$  and  $z \in \mathcal{Z}$ . In other words, we introduce a dummy RV  $T \in \mathcal{T}$  such that  $T$ ,  $(U, X)$ , and  $(Y, Z)$  form a Markov chain in this order, i.e.,  $T \rightarrow (U, X) \rightarrow (Y, Z)$ .

The idea of superposition coding is described as follows. The encoder  $g_n$  first maps the source message  $\mathbf{s}$  to a pair of  $n$ -length sequences  $(\mathbf{t}, \mathbf{u}) \in \mathcal{T}^n \times \mathcal{U}^n$  with a fixed type, say  $P_{TU}$ , and then sends the codeword  $\mathbf{u}$  over the channel, i.e.,  $g_n(\mathbf{s}) = \mathbf{u}$ . The encoder  $f_n$  first maps each pair  $(\mathbf{s}, \mathbf{l})$  to a triple of sequences  $(\mathbf{t}, \mathbf{u}, \mathbf{x}) \in \mathcal{T}^n \times \mathcal{U}^n \times \mathcal{X}^n$  such that  $\mathbf{x} \in \mathbb{T}_{P_{X|TU}}(\mathbf{t}, \mathbf{u})$ , then  $f_n$  sends the codeword  $\mathbf{x}$  over the channel, i.e.,  $f_n(\mathbf{s}, \mathbf{l}) = \mathbf{x}$ . In other words,  $g_n$  and  $f_n$  map  $(\mathbf{s}, \mathbf{l})$  to a tuple of sequences  $(\mathbf{t}, \mathbf{u}, \mathbf{x})$  with a joint type  $P_{TU}P_{X|TU}$ , although only  $\mathbf{u}$  and  $\mathbf{x}$  are sent to the channel, where  $\mathbf{t}$  plays the role of a dummy codeword.

Since  $W_{YZ|TUX}(\mathbf{y}, \mathbf{z}|\mathbf{t}, \mathbf{u}, \mathbf{x})$  is equal to  $W_{YZ|UX}(\mathbf{y}, \mathbf{z}|\mathbf{u}, \mathbf{x})$  and is independent of  $\mathbf{t}$ , transmitting the codewords  $(\mathbf{u}, \mathbf{x})$  through the channel  $W_{YZ|UX}$  can be viewed as transmitting the codewords  $(\mathbf{t}, \mathbf{u}, \mathbf{x})$  over the augmented channel  $W_{YZ|TUX}$ . Here, the common outputs of  $g_n$  and  $f_n$ ,  $(\mathbf{t}, \mathbf{u})$ 's, are called *auxiliary cloud centers* according to the traditional superposition coding notion [3], which convey the information of the common message  $\mathbf{s}$ , and the codewords  $\mathbf{x}$ 's corresponding to the same  $(\mathbf{t}, \mathbf{u})$  are called *satellite codewords* of  $(\mathbf{t}, \mathbf{u})$ , which contain both the common and private information, see Fig. 3. At the decoding stage, Receiver  $Z$  only needs to figure out which cloud  $(\mathbf{t}, \mathbf{u})$  was transmitted, and Receiver  $Y$  needs to estimate not only the cloud but also the satellite codeword  $\mathbf{x}$ . The introduction of the auxiliary RV  $T$  is made to enlarge the channel input alphabet from  $\mathcal{U} \times \mathcal{X}$  to  $\mathcal{T} \times \mathcal{U} \times \mathcal{X}$ , and the use of the superposition codeword  $\mathbf{t}$  renders the cloud centers  $(\mathbf{t}, \mathbf{u})$  more distinguishable by both receivers. We next employ superposition encoding to derive the achievable error exponent pair and the lower bound of system JSCC error exponent.

## 4.3 Achievable Exponents and a Lower Bound for $E_J$

Given arbitrary and finite alphabet  $\mathcal{T}$ , for any joint distribution  $P_{TUX} \in \mathcal{P}(\mathcal{T} \times \mathcal{U} \times \mathcal{X})$  and every  $R_1 > 0$ ,  $R_2 > 0$ , define

$$E_Y(R_1, R_2, W_{Y|TUX}, P_{TUX}) \triangleq \min_{V_{Y|TUX}} \left[ D(V_{Y|TUX} \| W_{Y|TUX} | P_{TUX}) \right]$$

$$+ \min \left( \left| I_{P_{TUX}V_{Y|TUX}}(T, U, X; Y) - (R_1 + R_2) \right|^+, \left| I_{P_{TUX}V_{Y|TUX}}(X; Y|T, U) - R_2 \right|^+ \right), \quad (17)$$

and

$$E_Z(R_1, R_2, W_{Z|TUX}, P_{TUX}) \triangleq \min_{V_{Z|TUX}} \left[ D(V_{Z|TUX} \| W_{Z|TUX} | P_{TUX}) + \left| I_{P_{TUX}V_{Z|TUX}}(T, U; Z) - R_1 \right|^+ \right], \quad (18)$$

where  $|x|^+ = \max(0, x)$ , and the outer minimum in (17) (respectively (18)) is taken over all conditional distributions on  $\mathcal{P}(\mathcal{Y}|\mathcal{T} \times \mathcal{U} \times \mathcal{X})$  (respectively  $\mathcal{P}(\mathcal{Z}|\mathcal{T} \times \mathcal{U} \times \mathcal{X})$ ). It immediately follows by definition that  $E_Y(R_1, R_2, W_{Y|TUX}, P_{TUX})$  is zero if and only if at least one of the following is satisfied

$$R_1 + R_2 \geq I_{P_{TUX}W_{Y|TUX}}(T, U, X; Y) = I(U, X; Y), \quad (19)$$

$$R_2 \geq I_{P_{TUX}W_{Y|TUX}}(X; Y|T, U), \quad (20)$$

and  $E_Z(R_1, R_2, W_{Z|TUX}, P_{TUX})$  is zero if and only if

$$R_1 \geq I_{P_{TUX}W_{Z|TUX}}(T, U; Z). \quad (21)$$

Using Lemma 3 and employing generalized maximum mutual information decoders at the two receivers, we can prove the following auxiliary bounds.

**Theorem 1** Given finite sets  $\mathcal{T}, \mathcal{U}, \mathcal{X}, \mathcal{Y}, \mathcal{Z}$ , a sequence of positive integers  $\{m_n\}$ , and a sequence of positive integers  $\{m'_{in}\}$  associated with every  $i = 1, 2, \dots, m_n$  with

$$\frac{1}{n} \log_2 m_n \rightarrow 0 \quad \text{and} \quad \frac{1}{n} \log_2 \max_i m'_{in} \rightarrow 0,$$

for any  $\delta > 0$ ,  $n$  sufficiently large, arbitrary (not necessarily distinct) types  $P_{(TU)_i} \in \mathcal{P}_n(\mathcal{T} \times \mathcal{U})$  and conditional types  $P_{X_j|(TU)_i} \in \mathcal{P}_n(\mathcal{X}|P_{(TU)_i})$ , and positive integers  $N_i$  and  $M_{ij}$ ,  $i = 1, 2, \dots, m_n$  and  $j = j(i) = 1, 2, \dots, m'_{in}$  with  $R_i < H_{P_{(TU)_i}}(T, U) - \delta$  and  $R_{ij} < H_{P_{(TU)_i}P_{X_j|(TU)_i}}(X|T, U) - \delta$ , where  $R_i \triangleq \frac{1}{n} \log_2 N_i$  and  $R_{ij} \triangleq \frac{1}{n} \log_2 M_{ij}$ , there exist  $m_n$  disjoint subsets  $\Omega_i = \left\{ (\mathbf{t}, \mathbf{u})_p^{(i)} \right\}_{p=1}^{N_i} \subseteq \mathbb{T}_{(TU)_i}$ ,  $m'_{in}$  disjoint subsets

$$\Omega_{ij}((\mathbf{t}, \mathbf{u})_p^{(i)}) = \left\{ \left( (\mathbf{t}, \mathbf{u})_p^{(i)}, \mathbf{x}_{p,q}^{(j)} \right) \right\}_{q=1}^{M_{ij}}$$

with  $\mathbf{x}_{p,q}^{(j)} \in \mathbb{T}_{X_j|(TU)_i}((\mathbf{t}, \mathbf{u})_p^{(i)})$  for every  $(\mathbf{t}, \mathbf{u})_p^{(i)} \in \Omega_i$  and every  $i$ , and a pair of mappings (decoding functions)  $\varphi_n^{(0)} : \mathcal{Y}^n \rightarrow \Omega$  and  $\psi_n^{(0)} : \mathcal{Z}^n \rightarrow \Omega$ , where  $\Omega \triangleq \bigcup_{ij} \Omega_{ij}$ , where  $\Omega_{ij} = \bigcup_{p=1}^{N_i} \Omega_{ij}((\mathbf{t}, \mathbf{u})_p^{(i)})$ , such that the probabilities of erroneous transmission of a triplet  $(\mathbf{t}, \mathbf{u}, \mathbf{x}) \in \Omega$  over the augmented channel  $W_{YZ|TUX}$  using decoders  $(\varphi_n^{(0)}, \psi_n^{(0)})$  are *simultaneously* bounded by

$$\begin{aligned} P_{Y_e}^{(n)}(\mathbf{t}, \mathbf{u}, \mathbf{x}) &\triangleq \sum_{\mathbf{y}: \varphi_n^{(0)}(\mathbf{y}) \neq (\mathbf{t}, \mathbf{u}, \mathbf{x})} W_{Y|TUX}^{(n)}(\mathbf{y}|\mathbf{t}, \mathbf{u}, \mathbf{x}) \\ &\leq 2^{-n} \left[ E_Y(R_i, R_{ij}, W_{Y|TUX}, P_{(TU)_i}, P_{X_j|(TU)_i}) - \delta \right] \end{aligned} \quad (22)$$

and

$$\begin{aligned}
P_{Z_e}^{(n)}(\mathbf{t}, \mathbf{u}, \mathbf{x}) &\triangleq \sum_{\mathbf{z}: \psi_n^{(0)}(\mathbf{z}) = ((\mathbf{t}, \mathbf{u})', \mathbf{x}') \text{ such that } (\mathbf{t}, \mathbf{u})' \neq (\mathbf{t}, \mathbf{u})} W_{Z|TUX}^{(n)}(\mathbf{z}|\mathbf{t}, \mathbf{u}, \mathbf{x}) \\
&\leq 2^{-n} \left[ E_Z \left( R_i, R_{ij}, W_{Z|TUX}, P_{(TU)_i}, P_{X_j|(TU)_i} \right) - \delta \right]
\end{aligned} \tag{23}$$

if  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}$  for every  $i, j$ .

**Proof:** We apply the packing lemma (Lemma 3) and a generalized MMI decoding rule.<sup>1</sup> In the sequel of the proof, we look at the superletter  $(T, U)$  (respectively  $X$ ) as the RV  $A$  (respectively  $B$ ) in Lemma 3. For the  $\{m_n\}$ ,  $\{m'_{in}\}$ ,  $P_{(TU)_i}$ ,  $P_{X_j|(TU)_i}$  given in Theorem 1, according to Lemma 3, there exist pairwise disjoint subsets  $\Omega_i$  and  $\Omega_{ij}((\mathbf{t}, \mathbf{u})_p^{(i)})$  satisfying (5), (6), and (7) for every  $1 \leq i \leq m_n$ ,  $1 \leq j \leq m'_{in}$ ,  $1 \leq p \leq N_i$ ,  $V_{(TU)'|TU} \in \mathcal{P}_n(\mathcal{T} \times \mathcal{U}|\mathcal{T} \times \mathcal{U})$ , and  $V_{(TU)'X'|TUX} \in \mathcal{P}_n(\mathcal{T} \times \mathcal{U} \times \mathcal{X}|\mathcal{T} \times \mathcal{U} \times \mathcal{X})$ , with the exception of the two cases that  $i = k$  and  $V_{(TU)'|TU}$  is the conditional distribution such that  $V_{(TU)'|TU}((t, u)'|(t, u))$  is 1 if  $(t, u)' = (t, u)$  and 0 otherwise, and that  $i = k, j = l$  and  $V_{(TU)'X'|TUX}$  is the conditional distribution such that  $V_{(TU)'X'|TUX}((t, u)', x'|t, u, x)$  is 1 if  $(t, u)' = (t, u), x' = x$  and 0 otherwise. Let

$$\Omega_{ij} = \bigcup_{p=1}^{N_i} \Omega_{ij}((\mathbf{t}, \mathbf{u})_p^{(i)}) \quad \text{and} \quad \Omega = \bigcup_{ij} \Omega_{ij}.$$

We shall show that for such  $\Omega_{ij}$ , there exists a pair of mappings  $(\varphi_n^{(0)}, \psi_n^{(0)})$  such that (22) and (23) are satisfied.

We first show that there exists a  $Y$ -decoder  $\varphi_n^{(0)}$  such that (22) holds. For any  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega$  and  $\mathbf{y} \in \mathcal{Y}^n$ , let

$$\alpha((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{y}) \triangleq I((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{y}) - (R_i + R_{ij}),$$

where  $R_i = \frac{1}{n} \log_2 N_i$  and  $R_{ij} = \frac{1}{n} \log_2 M_{ij}$  if  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}$ . Define  $Y$ -decoder  $\varphi_n^{(0)}: \mathcal{Y}^n \rightarrow \Omega$  by

$$\varphi_n^{(0)}(\mathbf{y}) \triangleq \arg \max_{((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega} \alpha((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{y}).$$

Using the decoder  $\varphi_n^{(0)}$ , we can upper bound the probability of error (assuming that  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}$  is sent through the channel) as follows

$$\begin{aligned}
P_{Y_e}^{(n)}((\mathbf{t}, \mathbf{u}), \mathbf{x}) &= W_{Y|TUX}^{(n)} \left( \left\{ \mathbf{y} : \varphi_n^{(0)}(\mathbf{y}) \neq ((\mathbf{t}, \mathbf{u}), \mathbf{x}) \right\} \middle| (\mathbf{t}, \mathbf{u}), \mathbf{x} \right) \\
&\leq \sum_{\hat{V}_{Y|TUX} \in \mathcal{P}_n(\mathcal{Y}|P_{(TU)_i X_j})} W_{Y|TUX}^{(n)} \left( \mathbb{T}_{\hat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \left\{ \mathbf{y} : \varphi_n^{(0)}(\mathbf{y}) \neq ((\mathbf{t}, \mathbf{u}), \mathbf{x}) \right\} \middle| \mathbf{t}, \mathbf{u}, \mathbf{x} \right). \tag{24}
\end{aligned}$$

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<sup>1</sup>Note that for the symmetric multiple access channel, it has been shown in [24] that the minimum conditional entropy (MCE) decoder leads to a larger channel error exponent than the MMI decoder; however, for the asymmetric 2-user channel with superposition coding, MMI decoding is equivalent to MCE decoding.

For any particular  $\widehat{V}_{Y|TUX}$ , since

$$\left\{ \mathbf{y} : \varphi_n^{(0)}(\mathbf{y}) \neq ((\mathbf{t}, \mathbf{u}), \mathbf{x}) \right\} = \underbrace{\left\{ \mathbf{y} : \varphi_n^{(0)}(\mathbf{y}) = ((\mathbf{t}, \mathbf{u})', \mathbf{x}'), (\mathbf{t}, \mathbf{u})' \neq (\mathbf{t}, \mathbf{u}) \right\}}_{\triangleq \mathcal{E}_1} \cup \underbrace{\left\{ \mathbf{y} : \varphi_n^{(0)}(\mathbf{y}) = ((\mathbf{t}, \mathbf{u}), \mathbf{x}'), \mathbf{x}' \neq \mathbf{x} \right\}}_{\triangleq \mathcal{E}_2},$$

we can upper bound

$$\begin{aligned} & W_{Y|TUX}^{(n)} \left( \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \left\{ \mathbf{y} : \varphi_n^{(0)}(\mathbf{y}) \neq ((\mathbf{t}, \mathbf{u}), \mathbf{x}) \right\} \middle| \mathbf{t}, \mathbf{u}, \mathbf{x} \right) \\ & \leq \sum_{\mathbf{y} \in \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_1} W_{Y|TUX}^{(n)}(\mathbf{y} | \mathbf{t}, \mathbf{u}, \mathbf{x}) + \sum_{\mathbf{y} \in \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_2} W_{Y|TUX}^{(n)}(\mathbf{y} | \mathbf{t}, \mathbf{u}, \mathbf{x}). \end{aligned} \quad (25)$$

It can be shown by the type packing lemma (Lemma 3) and a standard counting argument (see Appendix B) that

$$\begin{aligned} & \left| \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_1 \right| \leq m_n \left( \max_i m'_{in} \right) (n+1)^{|T \times \mathcal{U}|^2 |\mathcal{X}|^2 |\mathcal{Y}|} \\ & \quad \times 2^n \left[ H_{P_{(TU)_i X_j} \widehat{V}_{Y|TUX}}(Y|T, U, X) - \left| I_{P_{(TU)_i X_j} \widehat{V}_{Y|TUX}}(T, U, X; Y) - (R_i + R_{ij}) \right|^+ \right], \end{aligned} \quad (26)$$

and

$$\begin{aligned} & \left| \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_2 \right| \leq \left( \max_i m'_{in} \right) (n+1)^{|T \times \mathcal{U}|^2 |\mathcal{X}|^2 |\mathcal{Y}|} \\ & \quad \times 2^n \left[ H_{P_{(TU)_i X_j} \widehat{V}_{Y|TUX}}(Y|T, U, X) - \left| I_{P_{(TU)_i X_j} \widehat{V}_{Y|TUX}}(X; Y|T, U) - R_{ij} \right|^+ \right]. \end{aligned} \quad (27)$$

Using the identity (cf. Lemma 1) when  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij} \subseteq \mathbb{T}_{(TU)_i X_j}$  and  $\mathbf{y} \in \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x})$

$$W_{Y|TUX}^{(n)}(\mathbf{y} | (\mathbf{t}, \mathbf{u}), \mathbf{x}) = 2^{-n} \left[ D(\widehat{V}_{Y|TUX} \| W_{Y|TUX} | P_{(TU)_i X_j}) + H_{P_{(TU)_i X_j} \widehat{V}_{Y|TUX}}(Y|T, U, X) \right],$$

we obtain

$$\begin{aligned} & \sum_{\mathbf{y} \in \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_1} W_{Y|TUX}^{(n)}(\mathbf{y} | ((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}) \leq m_n \left( \max_i m'_{in} \right) (n+1)^{|T \times \mathcal{U}|^2 |\mathcal{X}|^2 |\mathcal{Y}|} \\ & \quad \times 2^{-n} \left[ D(\widehat{V}_{Y|TUX} \| W_{Y|TUX} | P_{(TU)_i X_j}) + \left| I_{P_{(TU)_i X_j} \widehat{V}_{Y|TUX}}(T, U, X; Y) - (R_i + R_{ij}) \right|^+ \right], \end{aligned} \quad (28)$$

and

$$\begin{aligned} & \sum_{\mathbf{y} \in \mathbb{T}_{\widehat{V}_{Y|TUX}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_2} W_{Y|TUX}^{(n)}(\mathbf{y} | ((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}) \leq \left( \max_i m'_{in} \right) (n+1)^{|T \times \mathcal{U}|^2 |\mathcal{X}|^2 |\mathcal{Y}|} \\ & \quad \times 2^{-n} \left[ D(\widehat{V}_{Y|TUX} \| W_{Y|TUX} | P_{(TU)_i X_j}) + \left| I_{P_{(TU)_i X_j} \widehat{V}_{Y|TUX}}(X; Y|T, U) - R_{ij} \right|^+ \right]. \end{aligned} \quad (29)$$

Substituting (28) and (29) back into (25) and (24) successively, noting that  $|\mathcal{P}_n(\mathcal{Y}|P_{(TU)_i X_j})|$  is polynomial in  $n$  by Lemma 1, we obtain that, for any  $\delta > 0$ , there exists a  $Y$ -decoder  $\varphi_n^{(0)}$  such that, given  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}$ , the probability of  $Y$ -error is bounded by

$$P_{Y_e}^{(n)}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \leq 2^{-n} \left[ E_Y \left( R_i, R_{ij}, W_{Y|TUX}, P_{(TU)_i} P_{X_j|(TU)_i} \right) - \delta \right] \quad (30)$$

for sufficiently large  $n$ .

Similarly, we can design a decoder for Receiver  $Z$  as follows. For any  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega$  and  $\mathbf{z} \in \mathcal{Z}^n$ , let

$$\beta((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{z}) = \beta((\mathbf{t}, \mathbf{u}); \mathbf{z}) \triangleq I((\mathbf{t}, \mathbf{u}); \mathbf{z}) - R_i,$$

where  $R_i = \frac{1}{n} \log_2 N_i$  if  $(\mathbf{t}, \mathbf{u}) \in \Omega_i$ . Note that  $\beta((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{z})$  is independent of  $\mathbf{x}$ . Let  $\tilde{\Omega} = \sum_{i=1}^{m_n} \Omega_i$ . The  $Z$ -decoder  $\psi_n^{(0)} : \mathcal{Z}^n \rightarrow \Omega$  is defined by

$$\begin{aligned} \varphi_n^{(0)}(\mathbf{z}) &= \arg \max_{((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega} \beta((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{z}) \\ &= ((\mathbf{t}, \mathbf{u})', \mathbf{x}') \quad \text{such that} \begin{cases} (\mathbf{t}, \mathbf{u})' = \arg \max_{(\mathbf{t}, \mathbf{u}) \in \tilde{\Omega}} \beta((\mathbf{t}, \mathbf{u}); \mathbf{z}), \\ \mathbf{x}' \text{ is arbitrary.} \end{cases} \end{aligned}$$

It can be shown in a similar manner by using (5) in Lemma 3 that, under the decoder  $\psi_n^{(0)}$ , the probability of the  $Z$ -error is bounded by

$$P_{Z_e}^{(n)}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \leq 2^{-n} \left[ E_Z \left( R_i, R_{ij}, W_{Z|TUX}, P_{(TU)_i} P_{X_j|(TU)_i} \right) - \delta \right] \quad (31)$$

for sufficiently large  $n$ . Finally, we remark that Lemma 3 ensures that there exist mappings  $(\varphi_n^{(0)}, \psi_n^{(0)})$  such that (31) holds simultaneously with (30).  $\blacksquare$

Theorem 1 is an auxiliary result for the channel coding problem for the 2-user asymmetric channel. To apply it to our 2-user source-channel system, we need to design encoders which can map a pair of correlated source messages to a particular  $(\mathbf{t}, \mathbf{u}, \mathbf{x})$  with a joint type, so that the total probabilities of error still vanish exponentially. We hence can establish the following bounds.

**Theorem 2** Given an arbitrary and finite alphabet  $\mathcal{T}$ , for any  $\tilde{P}_{TUX} \in \mathcal{P}(\mathcal{T} \times \mathcal{U} \times \mathcal{X})$ , the following exponent pair is universally achievable,

$$E_{JY}(Q_{SL}, W_{YZ|TUX}, \tilde{P}_{TUX}, \tau) \triangleq \min_{P_{SL}} \left[ \tau D(P_{SL} \| Q_{SL}) + E_Y(\tau H_P(S), \tau H_P(L|S), W_{Y|TUX}, \tilde{P}_{TUX}) \right], \quad (32)$$

and

$$E_{JZ}(Q_{SL}, W_{YZ|TUX}, \tilde{P}_{TUX}, \tau) \triangleq \min_{P_{SL}} \left[ \tau D(P_{SL} \| Q_{SL}) + E_Z(\tau H_P(S), \tau H_P(L|S), W_{Z|TUX}, \tilde{P}_{TUX}) \right], \quad (33)$$

where  $W_{Y|TUX}$  and  $W_{Z|TUX}$  are marginal distributions of  $W_{YZ|TUX}$ , which is the augmented conditional distribution from  $W_{YZ|UX}$ . Furthermore, given  $Q_{SL}$ ,  $W_{YZ|UX}$ , and  $\tau$ , the system JSCC error exponent satisfies

$$E_J(Q_{SL}, W_{YZ|UX}, \tau) \geq \min_{P_{SL}} [\tau D(P_{SL} \| Q_{SL}) + E_r(\tau H_P(S), \tau H_P(L|S), W_{YZ|UX})] \quad (34)$$

where

$$E_r(R_1, R_2, W_{YZ|UX}) \triangleq \sup_{\mathcal{T}} \max_{P_{TUX}} E_r(R_1, R_2, W_{YZ|TUX}, P_{TUX}), \quad (35)$$

where the supremum is taken over all finite alphabets  $\mathcal{T}$ , and the maximum is taken over all the joint distributions on  $\mathcal{P}(\mathcal{T} \times \mathcal{U} \times \mathcal{X})$  and  $E_r(R_1, R_2, W_{YZ|TUX}, P_{TUX})$  is given by

$$\min \{E_Y(R_1, R_2, W_{Y|TUX}, P_{TUX}), E_Z(R_1, R_2, W_{Z|TUX}, P_{TUX})\},$$

where  $E_Y$  and  $E_Z$  are given by (17) and (18), respectively.

We remark that (32) and (33) can be achieved by a sequence of codes without the knowledge of  $Q_{SL}$  and  $W_{YZ|UX}$ , but the lower bound (34) is achieved by a sequence of codes that needs to know the statistics of the channel.

**Proof of Theorem 2:** We first prove the achievable error exponent pair (32) and (33). We need to show that, for any given  $\tilde{P}_{TUX} \in \mathcal{P}(\mathcal{T} \times \mathcal{U} \times \mathcal{X})$  and  $\delta > 0$ , there exists a sequence of JSC codes such that both the probabilities of decoding error are upper bounded by

$$P_{ke}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \leq 2^{-n[E_{Jk}(Q_{SL}, W_{YZ|TUX}, \tilde{P}_{TUX}, \tau) - \delta]}, \quad k = Y, Z,$$

where  $E_{JY}$  and  $E_{JZ}$  are given by (32) and (33).

To apply Theorem 1, set  $m_n \triangleq |\mathcal{P}_{\tau n}(\mathcal{S})|$ . For each type  $P_{S_i} \in \mathcal{P}_{\tau n}(\mathcal{S})$ ,  $i = 1, 2, \dots, m_n$ , denote  $N_i$  be the cardinalities of these type classes,  $N_i \triangleq |\mathbb{T}_{S_i}|$ , and set  $m'_{in} \triangleq |\mathcal{P}_{\tau n}(\mathcal{L}|P_{S_i})|$ . For each conditional type  $P_{L_j|S_i} \in \mathcal{P}_{\tau n}(\mathcal{L}|P_{S_i})$ ,  $j = 1, 2, \dots, m'_{in}$ , denote  $M_{ij}$  be the cardinalities of these type classes,  $M_{ij} \triangleq |\mathbb{T}_{L_j|S_i}(\mathbf{s})|$  where  $\mathbf{s}$  is an arbitrary sequence in  $\mathbb{T}_{S_i}$ . Note that  $|\mathbb{T}_{L_j|S_i}(\mathbf{s})|$  is constant for all  $\mathbf{s} \in \mathbb{T}_{S_i}$ .  $R_i$  and  $R_{ij}$  are respectively given by  $\frac{1}{n} \log_2 N_i$  and  $\frac{1}{n} \log_2 M_{ij}$ .

Now no matter whether the given  $\tilde{P}_{TUX}$  belongs to  $\mathcal{P}_n(\mathcal{T} \times \mathcal{U} \times \mathcal{X})$  or not, we always can find a sequence of joint types  $\{P_{TUX} \in \mathcal{P}_n(\mathcal{T} \times \mathcal{U} \times \mathcal{X})\}_{n=1}^{\infty}$  such that  $P_{TUX} \rightarrow \tilde{P}_{TUX}$  uniformly<sup>2</sup> as  $n \rightarrow \infty$ . Thus, we can choose, by the continuity of  $E_k(R_i, R_{ij}, W_{k|TUX}, \tilde{P}_{TUX})$  with respect to  $\tilde{P}_{TUX}$ , for each  $i = 1, 2, \dots, m_n$ , and  $j = j(i) = 1, 2, \dots, m'_{in}$ , the joint type  $P_{(TU)_i X_j} = P_{TUX}$  such that the following are satisfied

$$\left| E_k(R_i, R_{ij}, W_{k|TUX}, P_{TUX}) - E_k(R_i, R_{ij}, W_{k|TUX}, \tilde{P}_{TUX}) \right| < \frac{\delta}{4}, \quad k = Y, Z$$

---

<sup>2</sup>We say that a sequence of distributions  $\{P_{X_i} \in \mathcal{P}(\mathcal{X})\}_{i=1}^{\infty}$  uniformly converges to  $P_X^* \in \mathcal{P}(\mathcal{X})$  if the variational distance [10] between  $P_{X_i}$  and  $P_X^*$  converges to zero as  $n \rightarrow \infty$ .

for  $n$  sufficiently large. Since the type  $P_{TUX}$  can also be regarded as a joint distribution, let  $P_{(TU)_i} = P_{TU} \in \mathcal{P}_n(\mathcal{T} \times \mathcal{U})$  be the marginal distribution on  $\mathcal{T} \times \mathcal{U}$  induced by  $P_{TUX}$  for all  $i = 1, 2, \dots, m_n$  and let  $P_{X_j|(TU)_i} = P_{X|TU} \in \mathcal{P}_n(\mathcal{X}|P_{TU})$  be the corresponding conditional distribution for all  $i = 1, 2, \dots, m_n$  and  $j = 1, 2, \dots, m'_{in}$ , i.e.,  $P_{X|TU}(\mathbf{x}|\mathbf{t}, \mathbf{u}) = P_{TUX}(\mathbf{t}, \mathbf{u}, \mathbf{x})/P_{TU}(\mathbf{t}, \mathbf{u})$  for any  $(\mathbf{t}, \mathbf{u}, \mathbf{x}) \in \mathbb{T}_{TUX}$ .

Without loss of generality, we assume, for the choice of  $N_i$ ,  $M_{ij}$ ,  $P_{(TU)_i}$ , and  $P_{X_j|(TU)_i}$ , the following conditions are satisfied for  $i = 1, 2, \dots, \widehat{m}_n$ ,  $j = 1, 2, \dots, \widehat{m}'_{in}$ ,

$$R_i < H_{P_{(TU)_i}}(T, U) - \frac{\delta}{4}, \quad i = 1, 2, \dots, \widehat{m}_n \quad (36)$$

and

$$R_{ij} < H_{P_{(TU)_i X_j}}(X|T, U) - \frac{\delta}{4}, \quad i = 1, 2, \dots, \widehat{m}_n, \quad j = j(i) = 1, 2, \dots, \widehat{m}'_{in}, \quad (37)$$

where  $\widehat{m}_n \leq m_n$  and  $\widehat{m}'_{in} \leq m'_{in}$ . Then according to Theorem 1, there exist pairwise disjoint subsets  $\Omega_{ij} \subseteq \mathbb{T}_{(TU)_i X_j}$  with  $|\Omega_{ij}| = N_i M_{ij}$ ,  $i = 1, 2, \dots, \widehat{m}_n$ ,  $j = 1, 2, \dots, \widehat{m}'_{in}$ , and a pair of mappings  $(\varphi_n^{(0)}, \psi_n^{(0)})$ , such that the probabilities of erroneous transmission of a  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}$  are *simultaneously* bounded for the channel  $W_{YZ|TUX}$  as

$$\begin{aligned} P_{Y_e}^{(n)}(\mathbf{t}, \mathbf{u}, \mathbf{x}) &\leq 2^{-n} [E_Y(R_i, R_{ij}, W_{Y|TUX}, P_{(TU)_i X_j}) - \delta/4] \\ &\leq 2^{-n} [E_Y(R_i, R_{ij}, W_{Y|TUX}, \tilde{P}_{TUX}) - \delta/2] \end{aligned} \quad (38)$$

and

$$\begin{aligned} P_{Z_e}^{(n)}(\mathbf{t}, \mathbf{u}, \mathbf{x}) &\leq 2^{-n} [E_Z(R_i, R_{ij}, W_{Z|TUX}, P_{(TU)_i X_j}) - \delta/4] \\ &\leq 2^{-n} [E_Z(R_i, R_{ij}, W_{Z|TUX}, \tilde{P}_{TUX}) - \delta/2]. \end{aligned} \quad (39)$$

For the  $N_i$ ,  $M_{ij}$ ,  $P_{(TU)_i}$ , and  $P_{X_j|(TU)_i}$  violating (36) or (37) (i.e., for  $i > \widehat{m}_n$  or  $j > \widehat{m}'_{in}$ ), (38) and (39) trivially hold for arbitrary choice of disjoint subsets  $\Omega_{ij}$  since  $E_Y(R_i, R_{ij}, W_{Y|TUX}, P_{(TU)_i X_j})$  or  $E_Z(R_i, R_{ij}, W_{Z|TUX}, P_{(TU)_i X_j})$  would be less than  $\delta/4$ . In fact, the functions  $E_Y$  and  $E_Z$  are trivially bounded by the following linear functions of  $R_i$  and  $R_{ij}$  with slope  $-1$  by definition,

$$\begin{aligned} E_Y(R_i, R_{ij}, W_{Y|TUX}, P_{(TU)_i X_j}) &\leq \min \left\{ I_{P_{(TU)_i X_j} W_{Y|TUX}}(T, U, X; Y) - R_i - R_{ij}, \right. \\ &\quad \left. I_{P_{(TU)_i X_j} W_{Y|TUX}}(X; Y|T, U) - R_{ij} \right\} \end{aligned} \quad (40)$$

and

$$E_Z(R_i, R_{ij}, W_{Z|TUX}, P_{(TU)_i X_j}) \leq I_{P_{(TU)_i X_j} W_{Z|TUX}}(T, U; Z) - R_i. \quad (41)$$

If  $R_i \geq H_{P_{(TU)_i}}(T, U) - \frac{\delta}{4} \geq I_{P_{(TU)_i X_j} W_{Z|TUX}}(T, U; Z) - \frac{\delta}{4}$ , then by (41)  $E_Z(R_i, R_{ij}, W_{Z|TUX}, P_{(TU)_i X_j}) \leq \frac{\delta}{4}$ . Similarly, if  $R_{ij} \geq H_{P_{(TU)_i X_j}}(X|T, U) - \frac{\delta}{4}$ , then by (40)  $E_Y(R_i, R_{ij}, W_{Y|TUX}, P_{(TU)_i X_j}) \leq \frac{\delta}{4}$ .

Therefore, we may construct the JSC code  $(f_n, g_n, \varphi_n, \psi_n)$  for CS  $Q_{SL}$  and the 2-user channel  $W_{YZ|UX}$  as follows. Without the loss of generality, we assume that the alphabets  $\mathcal{U}$  and  $\mathcal{X}$  contain the element 0.



Encoder  $g_n$ : For the message  $\mathbf{s} \in \mathbb{T}_{S_i}$  such that  $i > \widehat{m}_n$ , let  $g_n(\mathbf{s}) = \mathbf{0} \in \mathcal{U}^n$ . Denote  $\widetilde{\Omega} = \bigcup_i \Omega_i$ . For the  $\mathbf{s} \in \mathbb{T}_{S_i}$  such that  $i \leq \widehat{m}_n$ , let  $g_n^{(1)} : \mathcal{S}^{\tau n} \rightarrow \widetilde{\Omega}$  be a bijection that maps each  $\mathbf{s} \in \mathbb{T}_{S_i}$  to the corresponding  $(\mathbf{t}, \mathbf{u}) \in \Omega_i$ , by noting that  $|\Omega_i| = |\mathbb{T}_{S_i}| = N_i$ . Finally, let  $g_n(\mathbf{s})$  be the second component  $\mathbf{u}$  of  $g_n^{(1)}(\mathbf{s})$ .

Encoder  $f_n$ : For the message pair  $(\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{S_i L_j}$  such that  $i > \widehat{m}_n$  or  $j > \widehat{m}'_{in}$ , let  $f_n(\mathbf{s}, \mathbf{l}) = \mathbf{0} \in \mathcal{X}^n$ . For the  $(\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{S_i L_j}$  such that  $i \leq \widehat{m}_n$  and  $j \leq \widehat{m}'_{in}$ , noting that  $|\mathbb{T}_{L_j | S_i}(\mathbf{s})| = |\Omega_{ij}(\varphi_n(\mathbf{s}))| = M_{ij}$  if  $\mathbf{s} \in \mathbb{T}_{S_i}$ , let  $f_n^{(1)}(\mathbf{s}, \cdot) : \mathbb{T}_{L_j | S_i}(\mathbf{s}) \rightarrow \Omega_{ij}(g_n(\mathbf{s}))$  be a bijection such that  $f_n^{(1)}(\mathbf{s}, \mathbf{l}) = (g_n^{(1)}(\mathbf{s}), \mathbf{x}) \in \Omega_{ij}$ . Let  $f_n(\mathbf{s}, \mathbf{l})$  be the third component  $\mathbf{x}$  of  $f_n^{(1)}(\mathbf{s}, \mathbf{l})$ .

Clearly, the JSC encoders  $(f_n, g_n)$ , although working independently, they map each  $(\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{S_i L_j}$  to a unique pair  $(\mathbf{u}, \mathbf{x})$  when  $i \leq \widehat{m}_n$  and  $j \leq \widehat{m}'_{in}$ , and to  $(\cdot, \mathbf{0})$  otherwise (in this case an error is declared).

$Y$ -Decoder  $\varphi_n$ : The  $Y$ -decoder is defined by

$$\varphi_n(\mathbf{y}) \triangleq \begin{cases} (\mathbf{s}', \mathbf{l}') & \text{if } \exists (\mathbf{s}', \mathbf{l}') \in \mathcal{S}^n \times \mathcal{L}^n \text{ such that } f_n^{(1)}(\mathbf{s}', \mathbf{l}') = \varphi_n^{(0)}(\mathbf{y}), \\ (\mathbf{0}, \mathbf{0}) & \text{Otherwise.} \end{cases}$$

$Z$ -Decoder  $\psi_n$ : The  $Z$ -decoder is defined by

$$\psi_n(\mathbf{z}) \triangleq \begin{cases} \mathbf{s}' & \text{if } \exists \mathbf{s}' \in \mathcal{S}^n \text{ such that } g_n^{(1)}(\mathbf{s}') \text{ is equal to the first two components of } \psi_n^{(0)}(\mathbf{z}), \\ \mathbf{0} & \text{Otherwise.} \end{cases}$$

For such JSC code  $(f_n, g_n, \varphi_n, \psi_n)$ , the probabilities of  $Y$ -error and  $Z$ -error are bounded by

$$P_{Y_e}^{(n)}(\mathbf{s}, \mathbf{l}) \leq 2^{-n[E_Y(R_i, R_{ij}, W_Y | TUX, \widetilde{P}_{TUX}) - \delta/2]} \quad \text{if } (\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{S_i L_j} \quad (42)$$

and

$$P_{Z_e}^{(n)}(\mathbf{s}, \mathbf{l}) \leq 2^{-n[E_Z(R_i, R_{ij}, W_Z | TUX, \widetilde{P}_{TUX}) - \delta/2]} \quad \text{if } (\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{S_i L_j}. \quad (43)$$

Substituting (42) and (43) into (10) and using the fact (Lemma 1)  $Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) \leq 2^{-n\tau D(P_{SL} \| Q_{SL})}$ , we obtain, for  $n$  sufficiently large,

$$\begin{aligned} & P_{Y_e}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \\ & \leq \sum_{i,j} 2^{-n[\tau D(P_{S_i L_j} \| Q_{SL}) + E_Y(R_i, R_{ij}, W_Y | TUX, \widetilde{P}_{TUX}) - \delta/2]} \\ & \leq \sum_{P_{SL}} 2^{-n[\tau D(P_{SL} \| Q_{SL}) + E_Y(\tau H_P(S) - o_1(n), \tau H_P(L|S) - o_2(n), W_Y | TUX, \widetilde{P}_{TUX}) - \delta/2]} \\ & \leq \sum_{P_{SL}} 2^{-n[\tau D(P_{SL} \| Q_{SL}) + E_Y(\tau H_P(S), \tau H_P(L|S), W_Y | TUX, \widetilde{P}_{TUX}) - \delta]} \end{aligned} \quad (44)$$

and

$$\begin{aligned}
& P_{Z_e}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \\
& \leq \sum_{i,j} 2^{-n[\tau D(P_{S_i L_j} \| Q_{SL}) + E_Z(R_i, R_{ij}, W_{Z|TUX}, \tilde{P}_{TUX}) - \delta/2]} \\
& \leq \sum_{P_{SL}} 2^{-n[\tau D(P_{SL} \| Q_{SL}) + E_Z(\tau H_P(S) - o_1(n), \tau H_P(L|S) - o_2(n), W_{Z|TUX}, \tilde{P}_{TUX}) - \delta/2]} \\
& \leq \sum_{P_{SL}} 2^{-n[\tau D(P_{SL} \| Q_{SL}) + E_Z(\tau H_P(S), \tau H_P(L|S), W_{Z|TUX}, \tilde{P}_{TUX}) - \delta]}, \tag{45}
\end{aligned}$$

where  $o_1(n) = \frac{|\mathcal{S}| \log_2(\tau n + 1)}{n}$  and  $o_2(n) = \frac{|\mathcal{S}| |\mathcal{L}| \log_2(\tau n + 1)}{n}$ . Finally, the bounds (32) and (33) follow from (44) and (45), and the fact that the cardinality of set of joint types  $\mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})$  is upper bounded by  $(\tau n + 1)^{|\mathcal{S}| |\mathcal{L}|}$ .

To prove the lower bound (34), we slightly modify the above approach by choosing  $P_{(TU)_i X_j} = \tilde{P}_{(TU)_i X_j}^*$  which achieves the maximum and the supremum of  $E_r(R_i, R_{ij}, W_{YZ|UX})$  in (35) for every  $R_i$  and  $R_{ij}$ ,  $i = 1, 2, \dots, m_n$ ,  $j = 1, 2, \dots, m'_{in}$ . Then the probabilities of  $Y$ -error and  $Z$ -error in (42) and (43) are bounded by

$$\begin{aligned}
P_{Y_e}^{(n)}(\mathbf{s}, \mathbf{l}) & \leq 2^{-n[E_Y(R_i, R_{ij}, W_{YZ|UX}, \tilde{P}_{(TU)_i X_j}^*) - \delta/2]} \\
& \leq 2^{-n[E_r(R_i, R_{ij}, W_{YZ|UX}) - \delta/2]} \quad \text{if } (\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{S_i L_j} \tag{46}
\end{aligned}$$

and

$$\begin{aligned}
P_{Z_e}^{(n)}(\mathbf{s}, \mathbf{l}) & \leq 2^{-n[E_Z(R_i, R_{ij}, W_{Z|TUX}, \tilde{P}_{TUX}^*) - \delta/2]} \\
& \leq 2^{-n[E_r(R_i, R_{ij}, W_{YZ|UX}) - \delta/2]} \quad \text{if } (\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{S_i L_j} \tag{47}
\end{aligned}$$

for  $n$  sufficiently large. The rest of the proof is similar to the proofs of (32) and (33).  $\blacksquare$

By examining the positivity of the lower bound to  $E_J$ , we obtain a sufficient condition for reliable transmissibility for the asymmetric 2-user system. For the sake of completeness, we also prove a converse by using Fano's inequality, and hence establish the JSCC theorem for this system. Given  $W_{YZ|UX}$ , define

$$\mathcal{R}(W_{YZ|UX}) \triangleq \bigcup_{T: |T| \leq |\mathcal{U}| |\mathcal{X}| + 1} \bigcup_{P_{TUX} \in \mathcal{P}(T \times \mathcal{U} \times \mathcal{X})} \mathcal{R}(W_{YZ|TUX}, P_{TUX}) \tag{48}$$

where

$$\mathcal{R}(W_{YZ|TUX}, P_{TUX}) \triangleq \left\{ (R_1, R_2) : \begin{aligned} & R_1 + R_2 < I(T, U, X; Y) = I(U, X; Y) \\ & R_1 < I(T, U; Z) \\ & R_2 < I(X; Y|T, U) \end{aligned} \right\},$$

where the mutual informations are taken under the joint distribution  $P_{TUXYZ} = P_{TUX} W_{YZ|UX}$ . Note that  $\mathcal{R}(W_{YZ|UX})$  is convex and we denote  $\overline{\mathcal{R}}(W_{YZ|UX})$  be the closure of  $\mathcal{R}(W_{YZ|UX})$ .

**Theorem 3** (JSCC Theorem) Given  $Q_{SL}$ ,  $W_{YZ|UX}$  and  $\tau > 0$ , the following statements hold.

- (1) The sources  $Q_{SL}$  can be transmitted over the channel  $W_{YZ|UX}$  with probability of error  $P_e^{(n)} \rightarrow 0$  as  $n \rightarrow \infty$  if  $(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}(W_{YZ|UX})$ ;
- (2) Conversely, if the sources  $Q_{SL}$  can be transmitted over the channel  $W_{YZ|UX}$  with an arbitrarily small probability of error  $P_e^{(n)}$  as  $n \rightarrow \infty$ , then  $(\tau H_Q(S), \tau H_Q(L|S)) \in \overline{\mathcal{R}}(W_{YZ|UX})$ .

**Proof:** See Appendix C. ■

**Observation 1** Theorem 3 implies that  $\overline{\mathcal{R}}(W_{YZ|UX})$  is actually the capacity region for the asymmetric 2-user channel  $W_{YZ|UX}$ , as the JSCC reduces to the asymmetric 2-user channel coding if the sources come with a uniform joint distribution. It is shown in Appendix C that  $\overline{\mathcal{R}}(W_{YZ|UX})$  can be equivalently written as

$$\mathcal{R}'(W_{YZ|UX}) \triangleq \bigcup_{T:|T| \leq |\mathcal{U}|+1} \bigcup_{P_{TUX} \in \mathcal{P}(T \times \mathcal{U} \times \mathcal{X})} \mathcal{R}'(W_{YZ|TUX}, P_{TUX}) \quad (49)$$

where

$$\mathcal{R}'(W_{YZ|TUX}, P_{TUX}) \triangleq \left\{ (R_1, R_2) : \begin{array}{l} R_1 + R_2 \leq \min\{I(U, X; Y), I(T, U; Z) + I(X; Y|T, U)\} \\ R_1 \leq I(T, U; Z) \end{array} \right\}.$$

Recently, Liang *et al.* [23] also showed (using a different approach) that the capacity region for the asymmetric 2-user channel is given by

$$\mathcal{R}''(W_{YZ|UX}) \triangleq \bigcup_{T:|T| \leq |\mathcal{U}|+1} \bigcup_{P_{TUX} \in \mathcal{P}(T \times \mathcal{U} \times \mathcal{X})} \mathcal{R}''(W_{YZ|TUX}, P_{TUX}) \quad (50)$$

where

$$\mathcal{R}''(W_{YZ|TUX}, P_{TUX}) \triangleq \left\{ (R_1, R_2) : \begin{array}{l} R_1 + R_2 \leq \min\{I(U, X; Y), I(T, U; Z) + I(X; Y|T, U)\} \\ R_1 \leq I(T, U; Z) \\ R_2 \leq I(X; Y|U) \end{array} \right\},$$

where the mutual informations are taken under the joint distribution  $P_{TUXYZ} = P_{TUX}W_{YZ|UX}$ . They state that our capacity region,  $\overline{\mathcal{R}}(W_{YZ|UX})$ , is a subset of their region  $\mathcal{R}''(W_{YZ|UX})$  described above by (50); this holds since  $I(X; Y|T, U) \leq I(X; Y|U)$ . However this is only partially correct, since noting that  $\mathcal{R}''(W_{YZ|UX}) \subseteq \mathcal{R}'(W_{YZ|UX})$  and that  $\mathcal{R}'(W_{YZ|UX}) = \overline{\mathcal{R}}(W_{YZ|UX})$  (as shown in Appendix C), one directly obtains that  $\mathcal{R}''(W_{YZ|UX}) \subseteq \overline{\mathcal{R}}(W_{YZ|UX})$ . Thus the regions are all identical:  $\overline{\mathcal{R}}(W_{YZ|UX}) = \mathcal{R}'(W_{YZ|UX}) = \mathcal{R}''(W_{YZ|UX})$ .

#### 4.4 Separation Principle for the Asymmetric 2-User System

It can be verified that the condition  $(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}(W_{YZ|UX})$  of Theorem 3 can be achieved by separate source and channel coding. The separate coding system of rate  $\tau$  (source symbol/channel symbol) (we refer to it by the *tandem* coding system) is depicted in Figs. 4 and 5 (with  $\pi_f$  and  $\pi_g$  being identity mappings).

The encoder  $f_n$  is composed of two source encoders  $f_{sn} : \mathcal{L}^{\tau n} \rightarrow \{1, 2, \dots, M_l\}$  and  $g_{sn} : \mathcal{S}^{\tau n} \rightarrow \{1, 2, \dots, M_s\}$  with private source coding rate  $\widehat{R}_l \triangleq \frac{1}{\tau n} \log_2 M_l$  and common source coding rate  $\widehat{R}_s \triangleq \frac{1}{\tau n} \log_2 M_s$  and a channel encoder  $\{1, 2, \dots, M_l\} \times \{1, 2, \dots, M_s\} \rightarrow \mathcal{X}^n$ . Similarly, the encoder  $g_n$  is composed of a source encoder  $g_{sn} : \mathcal{S}^{\tau n} \rightarrow \{1, 2, \dots, M_s\}$  with common coding rate  $\widehat{R}_s$  and a channel encoder  $g_{cn} : \{1, 2, \dots, M_s\} \rightarrow \mathcal{U}^n$ .

At the receiver side, the decoder  $\varphi_n$  is composed of a channel decoder  $\varphi_{cn} : \mathcal{Y}^n \rightarrow \{1, 2, \dots, M_l\} \times \{1, 2, \dots, M_s\}$ , and a source decoder  $\varphi_{sn} : \{1, 2, \dots, M_l\} \times \{1, 2, \dots, M_s\} \rightarrow \mathcal{S}^{\tau n} \times \mathcal{L}^{\tau n}$  which outputs the approximation of the source messages  $\mathbf{s}'$  and  $\mathbf{l}'$ . Similarly, the decoder  $\psi_n$  is composed of a channel decoder  $\psi_{cn} : \mathcal{Z}^n \rightarrow \{1, 2, \dots, M_s\}$ , and a source decoder  $\psi_{sn} : \{1, 2, \dots, M_s\} \rightarrow \mathcal{S}^{\tau n}$ .

To show that the condition  $(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}(W_{YZ|UX})$  can be achieved by the above tandem system, we need to apply the following 2-user source and channel coding theorems (we only state the forward parts of the theorems). Note that both of these theorems are special case of Theorem 3.

Let  $(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})$  be a sequence of source codes for CS  $Q_{SL}$  with common source rate  $\widehat{R}_s$  and private source rate  $\widehat{R}_l$  as defined above. The probability of the overall 2-user source coding error is given by

$$P_{es}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \triangleq \Pr \left( \{\varphi_{sn}(g_{sn}(S^{\tau n}), f_{sn}(L^{\tau n})) \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_{sn}(g_{sn}(S^{\tau n})) \neq S^{\tau n}\} \right). \quad (51)$$

Then by the 2-user source coding theorem, there exists a sequence of source codes  $(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})$  with rates  $\widehat{R}_s$  and  $\widehat{R}_l$  such that  $P_{es}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \rightarrow 0$  as  $n \rightarrow \infty$  if the rates satisfy  $\widehat{R}_s > H_Q(S)$  and  $\widehat{R}_l > H_Q(L|S)$ , i.e.,  $(\widehat{R}_s, \widehat{R}_l)$  lies in the upper-right infinite rectangle with vertex given by the point  $(H_Q(S), H_Q(L|S))$ .

We next state the forward part of channel coding theorem for the asymmetric 2-user channel. Let the (common and private) message pair  $(j, i)$  be uniformly drawn from the finite set  $\mathcal{M}_s \times \mathcal{M}_l$ , where  $\mathcal{M}_s \triangleq \{1, 2, \dots, M_s\}$  and  $\mathcal{M}_l \triangleq \{1, 2, \dots, M_l\}$ , and let  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$  be an asymmetric 2-user channel code with block length  $n$  and common and private message sets  $\mathcal{M}_s$  and  $\mathcal{M}_l$ . Let  $R_s \triangleq \frac{1}{n} \log_2 M_s$  and  $R_l \triangleq \frac{1}{n} \log_2 M_l$  be the common and private rates of the channel code, respectively. The average probability of error for asymmetric 2-user channel coding is given by

$$P_{ec}^{(n)}(R_s, R_l, W_{YZ|UX}) \triangleq \Pr \left( \{\varphi_{cn}(Y^n) \neq (J, I)\} \cup \{\psi_{cn}(Z^n) \neq J\} \right), \quad (52)$$

where  $(J, I)$  are uniformly drawn from  $\mathcal{M}_s \times \mathcal{M}_l$ . The maximum probability for error of asymmetric 2-user

channel coding is given by

$$P_{ec,max}^{(n)}(R_s, R_l, W_{YZ|UX}) \triangleq \max_{(j,i) \in \mathcal{M}_s \times \mathcal{M}_l} \Pr \left( \{\varphi_{cn}(Y^n) \neq (J, I)\} \cup \{\psi_{cn}(Z^n) \neq J\} \mid J = j, I = i \right), \quad (53)$$

Then there exists a sequence of channel codes  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$  such that  $P_{ec}^{(n)}(R_s, R_l, W_{YZ|UX}) \rightarrow 0$  as  $n \rightarrow \infty$  if  $(R_s, R_l) \in \mathcal{R}(W_{YZ|UX})$ . Furthermore, it can be readily shown by a standard expurgation argument [6, p. 204] that  $P_{ec,max}^{(n)}(R_s, R_l, W_{YZ|UX}) \rightarrow 0$  as  $n \rightarrow \infty$  if  $(R_s, R_l) \in \mathcal{R}(W_{YZ|UX})$ .

Now by (14), the overall probability of error for the tandem system is given by

$$P_e^{(n)} \triangleq \Pr \left( \{\varphi_{sn}[\varphi_{cn}(Y^n)] \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_{sn}[\psi_{cn}(Z^n)] \neq S^{\tau n}\} \right).$$

By the union bound, it is easy to see that  $P_e^{(n)}$  is upper bounded by

$$\begin{aligned} P_e^{(n)} &\leq \Pr \left( \{\varphi_{sn}(g_{sn}(S^{\tau n}), f_{sn}(L^{\tau n})) \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_{sn}(g_{sn}(S^{\tau n})) \neq S^{\tau n}\} \right) \\ &+ \Pr \left( \{\varphi_{cn}(Y^n) \neq (g_{sn}(S^{\tau n}), f_{sn}(L^{\tau n}))\} \cup \{\psi_{cn}(Z^n) \neq g_{sn}(S^{\tau n})\} \right) \\ &= P_{es}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \\ &+ \sum_{(j,i) \in \mathcal{M}_s \times \mathcal{M}_l} \Pr(g_{sn}(S^{\tau n}) = j, f_{sn}(L^{\tau n}) = i) \Pr \left( \{\varphi_{cn}(Y^n) \neq (J, I)\} \cup \{\psi_{cn}(Z^n) \neq J\} \mid J = j, I = i \right) \\ &\leq P_{es}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) + P_{ec,max}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) \end{aligned}$$

where  $P_{ec,max}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX})$  is the maximum channel coding probability of error with common rate  $\tau \widehat{R}_s$  and private rate  $\tau \widehat{R}_l$ . Clearly, by combining the 2-user source coding theorem and the asymmetric 2-user channel coding theorem, if  $(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}(W_{YZ|UX})$ , then there exist a sequence of source codes  $(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})$  and a sequence of channel codes  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$  such that the overall tandem system probability of error  $P_e^{(n)} \rightarrow 0$  as  $n \rightarrow \infty$ . Therefore, separation of source and channel coding is optimal from the point of view of reliable transmissibility.

#### 4.5 The Upper Bound to $E_J$

In [8], Csiszár also established an upper bound for the JSCC error exponent for the point-to-point discrete memoryless source-channel system in terms of the source and channel error exponents by a simple type counting argument. He shows that the JSCC error exponent is always less than the infimum of the sum of the source and channel error exponent, even though the channel error exponent is only partially known for high rates. This conceptual bound cannot currently be computed as the channel error exponent is not yet fully known for all achievable coding rates, but it directly implies that any upper bound for the channel error exponent yields a corresponding upper bound for the JSCC error exponent. For the asymmetric 2-user channel, a similar bound can be shown.

**Definition 2** The asymmetric 2-user channel coding error exponent  $E(R_1, R_2, W_{YZ|UX})$ , for any  $R_1 > 0$  and  $R_2 > 0$ , is defined by the supremum of the set of all numbers  $E_c$  for which there exists a sequence of asymmetric channel codes  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$  with blocklength  $n$ , the common rate no less than  $R_1$ , and the private rate no less than  $R_2$ , such that

$$E_c \leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{ec}^{(n)}(R_1, R_2, W_{YZ|UX}). \quad (54)$$

Denote the probabilities of  $Y$ - and  $Z$ -error of the channel coding by

$$P_{Yec}^{(n)}(R_s, R_l, W_{YY|UX}) \triangleq \Pr(\{\varphi_{cn}(Y^n) \neq (J, I)\}) = \frac{1}{2^{R_1+R_2}} \sum_{\mathcal{M}_s \times \mathcal{M}_l} \sum_{\mathbf{y}: \varphi_{cn}(\mathbf{y}) \neq (j, i)} W_{Y|X}^{(n)}(\mathbf{y}|\mathbf{u}, \mathbf{x}) \quad (55)$$

and

$$P_{Zec}^{(n)}(R_s, R_l, W_{YZ|UX}) \triangleq \Pr(\{\psi_{cn}(Z^n) \neq J\}) = \frac{1}{2^{R_1+R_2}} \sum_{\mathcal{M}_s \times \mathcal{M}_l} \sum_{\mathbf{z}: \psi_{cn}(\mathbf{z}) \neq j} W_{Z|X}^{(n)}(\mathbf{z}|\mathbf{u}, \mathbf{x}) \quad (56)$$

where  $\mathbf{x} \triangleq f_{cn}(j, i)$  and  $\mathbf{u} \triangleq g_{cn}(j)$ . Clearly, for any sequence of channel codes  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$ ,  $P_e^{(n)}(R_1, R_2, W_{YZ|UX})$  must be larger than  $P_{Yec}^{(n)}(R_1, R_2, W_{Y|UX})$  and  $P_{Zec}^{(n)}(R_1, R_2, W_{Z|UX})$  but less than the sum of the two, so we have

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{ec}^{(n)}(R_1, R_2, W_{YZ|UX}) = \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \max \left( P_{Yec}^{(n)}(R_1, R_2, W_{Y|UX}), P_{Zec}^{(n)}(R_1, R_2, W_{Z|UX}) \right). \quad (57)$$

Our upper bound for the system JSCC error exponent  $E_J$  (defined in Definition 1) is stated as follows.

**Theorem 4** Given  $Q_{SL}$ ,  $W_{YZ|UX}$ , and  $\tau$ , the system JSCC error exponent satisfies

$$E_J(Q_{SL}, W_{YZ|UX}, \tau) \leq \inf_{P_{SL}} [\tau D(P_{SL} \| Q_{SL}) + E(\tau H_P(S), \tau H_P(L|S), W_{YZ|UX})], \quad (58)$$

where  $E(\cdot, \cdot, W_{YZ|UX})$  is the corresponding channel coding error exponent for the asymmetric 2-user channel as defined above in Definition 2.

**Proof:** First, from (10) we can write

$$P_{ie}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \geq \max_{P_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})} Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) P_{ie}(\mathbb{T}_{SL}) \quad i = Y, Z, \quad (59)$$

where  $P_{Ye}(\mathbb{T}_{SL})$  and  $P_{Ze}(\mathbb{T}_{SL})$  are given by (11) and (12), respectively. Comparing (11) with (55), and comparing (12) with (56), we note that  $P_{Ye}(\mathbb{T}_{SL})$  and  $P_{Ze}(\mathbb{T}_{SL})$  can be interpreted as the probabilities of  $Y$ -error and  $Z$ -error of the asymmetric 2-user channel coding with (common and private) message sets  $\mathbb{T}_{SL}$ , since  $(\mathbf{s}, \mathbf{l})$  are uniformly distributed on  $\mathbb{T}_{SL}$ . For any  $P_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})$ , let  $P_S$  and  $P_{L|S}$  be the marginal and conditional distributions induced by  $P_{SL}$ . Recall that for each  $\mathbf{s} \in \mathbb{T}_S = \mathbb{T}_{P_S}$ ,

$$\mathbb{T}_{L|S}(\mathbf{s}) \triangleq \mathbb{T}_{P_{L|S}}(\mathbf{s}) = \{\mathbf{l} : (\mathbf{s}, \mathbf{l}) \in \mathbb{T}_{SL}\}$$

and that  $\mathbb{T}_{L|S}(\mathbf{s})$  is the same set for all  $\mathbf{s} \in \mathbb{T}_S$ . Hence, we can write  $\mathbb{T}_{SL}$  by the product of two sets  $\mathbb{T}_{SL} = \mathbb{T}_S \times \mathbb{T}_{L|S}(\mathbf{s})$ . Setting  $\tilde{R}_1 = \frac{1}{n} \log_2 |\mathbb{T}_S|$  and  $\tilde{R}_2 = \frac{1}{n} \log_2 |\mathbb{T}_{L|S}(\mathbf{s})|$ , it follows that, by the definition of asymmetric 2-user channel coding error exponent and (57),

$$\begin{aligned} \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \max_{i=Y,Z} P_{ie}(\mathbb{T}_{SL}) &\leq E(\liminf_{n \rightarrow \infty} \tilde{R}_1, \liminf_{n \rightarrow \infty} \tilde{R}_2, W_{YZ|UX}) \\ &= E(\tau H_P(S), \tau H_P(L|S), W_{YZ|UX}) \end{aligned} \quad (60)$$

for any sequence of JSC codes  $(f_n, \varphi_n, \psi_n)$ , recalling from Lemma 1 that

$$(\tau n + 1)^{-|\mathcal{S}|} 2^{n\tau H_P(S)} \leq |\mathbb{T}_S| \leq 2^{n\tau H_P(S)}$$

and

$$(\tau n + 1)^{-|\mathcal{S}||\mathcal{L}|} 2^{n\tau H_P(L|S)} \leq |\mathbb{T}_{L|S}(\mathbf{s})| \leq 2^{n\tau H_P(L|S)}.$$

According to (16), we write

$$\begin{aligned} &\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_e^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \\ &= \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \max \left( P_{Ye}^{(n)}(Q_{SL}, W_{Y|X}, \tau), P_{Ze}^{(n)}(Q_{SL}, W_{Z|X}, \tau) \right) \\ &\leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \max_{i=Y,Z} \max_{P_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})} Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) P_{ie}(\mathbb{T}_{SL}) \\ &= \liminf_{n \rightarrow \infty} \min_{P_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})} -\frac{1}{n} \log_2 Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) \max_{i=Y,Z} P_{ie}(\mathbb{T}_{SL}) \\ &= \liminf_{n \rightarrow \infty} \min_{P_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})} \left[ -\frac{1}{n} \log_2 Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) - \frac{1}{n} \log_2 \max_{i=Y,Z} P_{ie}(\mathbb{T}_{SL}) \right]. \end{aligned} \quad (61)$$

By Lemma 1, for any  $P_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})$ ,

$$-\frac{1}{\tau n} \log_2 Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) \leq D(P_{SL} \| Q_{SL}) + |\mathcal{S}||\mathcal{L}| \frac{1}{\tau n} \log_2(1 + \tau n)$$

which implies

$$\limsup_{n \rightarrow \infty} -\frac{1}{n} \log_2 Q_{SL}^{(\tau n)}(\mathbb{T}_{SL}) \leq \tau D(P_{SL} \| Q_{SL}). \quad (62)$$

Now assume that

$$\inf_{P_{SL} \in \mathcal{P}(\mathcal{S} \times \mathcal{L})} \left[ \tau D(P_{SL} \| Q_{SL}) + E(\tau H_P(S), \tau H_P(L|S), W_{YZ|UX}) \right]$$

is finite (the upper bound is trivial if it is infinity) and the infimum actually becomes a minimum. Let the minimum be achieved by distribution  $P_{SL}^* \in \mathcal{P}(\mathcal{S} \times \mathcal{L})$ , then there must exist a sequence of types  $\{\hat{P}_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})\}_{n=n_o}^{\infty}$  such that  $\hat{P}_{SL} \rightarrow P_{SL}^*$  uniformly<sup>3</sup>. It then follows from (61), (60) and (62) that

$$\begin{aligned} &\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_e^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \\ &\leq \liminf_{n \rightarrow \infty} \left[ -\frac{1}{n} \log_2 Q_{SL}^{(\tau n)}(\mathbb{T}_{\hat{P}_{SL}}) - \frac{1}{n} \log_2 \max_{i=Y,Z} P_{ie}(\mathbb{T}_{\hat{P}_{SL}}) \right] \\ &\leq \tau D(P_{SL}^* \| Q_{SL}) + E(\tau H_{P^*}(S), \tau H_{P^*}(L|S), W_{YZ|UX}). \end{aligned} \quad (63)$$

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<sup>3</sup>The sequence  $\{\hat{P}_{SL} \in \mathcal{P}_{\tau n}(\mathcal{S} \times \mathcal{L})\}_{n=n_o}^{\infty}$  here denotes a sequence for  $n = n_o, 2n_o, 3n_o, \dots$ , where  $n_o$  is the smallest integer such that  $\tau n_o$  is also an integer.

Since the above bound holds for any sequence of JSC codes, we complete the proof of Theorem 4.  $\blacksquare$

## 5 Applications to CS-AMAC and CS-ABC Systems

As pointed out in the introduction, our results obtained in the previous section can be directly applied to the CS-AMAC and CS-ABC source-channel systems.

### 5.1 CS-AMAC System

Setting  $|\mathcal{Z}| = 1$  and removing the decoder  $\psi_n$ , the 2-user asymmetric channel  $W_{YZ|UX}$  reduces to an AMAC  $W_{Y|UX}$ . Since the CS-AMAC system is a special case of the 2-user system, the quantities defined before, including the system (overall) probability of error, the system JSCC error exponent, and the channel error exponent still hold for the CS-AMAC system. Note that there is only one decoder, so we do not have the  $Z$ -error probability (nor exponent) here. The first union in (48) can be removed since the largest region is given by  $|\mathcal{T}| = 1$ . In fact, for any  $T \rightarrow (U, X) \rightarrow Y$ ,  $I(T, U, X; Y) = I(U, X; Y)$  and  $I(X; Y|T, U) \leq I(X; Y|U)$ . Thus Theorem 3 reduces to the same JSCC theorem established in [4] for the CS-AMAC system. Choosing the auxiliary alphabet  $|\mathcal{T}| = 1$ , we specialize Theorems 2 and 4 to the following corollary.

**Corollary 1** Given  $Q_{SL}$ ,  $W_{Y|UX}$  and  $\tau$ , the system JSCC error exponent satisfies

$$E_J(Q_{SL}, W_{Y|UX}, \tau) \geq \min_{P_{SL}} [\tau D(P_{SL} \| Q_{SL}) + E_r(\tau H_P(S), \tau H_P(L|S), W_{Y|UX})], \quad (64)$$

and

$$E_J(Q_{SL}, W_{Y|UX}, \tau) \leq \inf_{P_{SL}} [\tau D(P_{SL} \| Q_{SL}) + E(\tau H_P(S), \tau H_P(L|S), W_{Y|UX})], \quad (65)$$

where  $E(\tau H_P(S), \tau H_P(L|S), W_{Y|UX})$  is the channel error exponent of the AMAC  $W_{Y|UX}$  defined in (54) with  $|\mathcal{Z}| = 1$ , and

$$E_r(R_1, R_2, W_{Y|UX}) = \max_{P_{UX}} E_Y(R_1, R_2, W_{Y|UX}, P_{UX}) \quad (66)$$

where  $E_Y(R_1, R_2, W_{Y|UX}, P_{UX})$  is defined in (17) with  $|\mathcal{T}| = 1$ .

It has been shown in [2] that for any  $R_1 > 0$  and  $R_2 > 0$ , the channel exponent for AMAC  $W_{Y|UX}$  satisfies

$$E(R_1, R_2, W_{YZ|X}) \leq E_{sp}(R_1, R_2, W_{Y|UX}),$$

where

$$E_{sp}(R_1, R_2, W_{Y|UX}) \triangleq \max_{P_{UX} \in \mathcal{P}(\mathcal{U} \times \mathcal{X})} \min D(V_{Y|UX} \| W_{Y|UX} | P_{UX}), \quad (67)$$

where the minimum is taken over  $V_{Y|UX} \in \mathcal{P}(\mathcal{Y}|\mathcal{U} \times \mathcal{X})$  such that  $I_{P_{UX}V_{Y|UX}}(U, X; Y) \leq R_1 + R_2$  or  $I_{P_{UX}V_{Y|UX}}(X; Y|U) \leq R_2$ .



As a consequence, we obtain that

$$E_J(Q_{SL}, W_{Y|UX}, \tau) \leq \inf_{P_{SL}} [\tau D(P_{SL} \parallel Q_{SL}) + E_{sp}(\tau H_P(S), \tau H_P(L|S), W_{Y|UX})]. \quad (68)$$

In Section 6, we investigate the evaluation of lower bound (64) and upper bound (68) when the AMAC has a symmetric distribution.

## 5.2 CS-ABC System

Setting  $|\mathcal{U}| = 1$  and removing the encoder  $g_n$ , the 2-user asymmetric channel  $W_{YZ|UX}$  reduces to an ABC  $W_{YZ|X}$ . The quantities defined before, including the probabilities of error at  $Y$ -decoder and  $Z$ -decoder, the achievable error exponent pair, system (overall) probability of error, the system JSCC error exponent, and the channel error exponent still hold for the CS-ABC system. Given an arbitrary and finite auxiliary alphabet  $\mathcal{T}$ , we augment the channel  $W_{YZ|X}$  to  $W_{YZ|TX}$  by introducing a RV  $T \in \mathcal{T}$  such that  $T \rightarrow X \rightarrow (YZ)$ . Similarly, the marginal distributions of the augmented channel are denoted by  $W_{Y|TX}$  and  $W_{Z|TX}$ . We then specialize Theorems 2, 3 and 4 to the following corollaries.

Given  $W_{YZ|X}$ ,  $\mathcal{R}(W_{YZ|UX})$  of (48) reduces to  $\mathcal{R}(W_{YZ|X})$  given by

$$\mathcal{R}(W_{YZ|X}) \triangleq \bigcup_{\mathcal{T}: |\mathcal{T}| \leq |\mathcal{X}|+1} \bigcup_{P_{TX} \in \mathcal{P}(\mathcal{T} \times \mathcal{X})} \mathcal{R}(W_{YZ|TX}, P_{TX}) \quad (69)$$

where

$$\mathcal{R}(W_{YZ|TX}, P_{TX}) = \left\{ (R_1, R_2) : \begin{array}{l} R_1 + R_2 < I(T, X; Y) = I(X; Y) \\ R_1 < I(T; Z) \\ R_2 < I(X; Y|T) \end{array} \right\},$$

where the mutual informations are taken under the joint distribution  $P_{TXYZ} = P_{TX} W_{YZ|X}$ . We remark that the closure of  $\mathcal{R}(W_{YZ|X})$ , denoted by  $\overline{\mathcal{R}}(W_{YZ|X})$ , is the capacity region of the ABC  $W_{YZ|X}$  [21].

**Corollary 2** (JSCC Theorem for CS-ABC system) Given  $Q_{SL}$ ,  $W_{YZ|X}$  and  $\tau > 0$ , the following statements hold.

(1) The sources  $Q_{SL}$  can be transmitted over the ABC  $W_{YZ|X}$  with  $P_e^{(n)} \rightarrow 0$  as  $n \rightarrow \infty$  if

$$(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}(W_{YZ|X});$$

(2) Conversely, if the sources  $Q_{SL}$  can be transmitted over the ABC  $W_{YZ|X}$  with an arbitrarily small probability of error  $P_e^{(n)}$  as  $n \rightarrow \infty$ , then  $(\tau H_Q(S), \tau H_Q(L|S)) \in \overline{\mathcal{R}}(W_{YZ|X})$ .

**Corollary 3** Given an arbitrary and finite alphabet  $\mathcal{T}$ , for any  $\tilde{P}_{TX} \in \mathcal{P}(\mathcal{T} \times \mathcal{X})$ , the following exponent pair is universally achievable,

$$E_{JY}(Q_{SL}, W_{YZ|TX}, \tilde{P}_{TX}, \tau) \triangleq \min_{P_{SL}} \left[ \tau D(P_{SL} \parallel Q_{SL}) + E_Y(\tau H_P(S), \tau H_P(L|S), W_{Y|TX}, \tilde{P}_{TX}) \right], \quad (70)$$

and

$$E_{JZ}(Q_{SL}, W_{YZ|TX}, \tilde{P}_{TX}, \tau) \triangleq \min_{P_{SL}} \left[ \tau D(P_{SL} \parallel Q_{SL}) + E_Z(\tau H_P(S), \tau H_P(L|S), W_{Z|TX}, \tilde{P}_{TX}) \right], \quad (71)$$

where  $E_Y$  and  $E_Z$  are defined in (17) and (18) by setting  $|\mathcal{U}| = 1$ . Furthermore, given  $Q_{SL}$ ,  $W_{YZ|X}$ , and  $\tau$ , the system JSCC error exponent satisfies

$$E_J(Q_{SL}, W_{YZ|X}, \tau) \geq \min_{P_{SL}} \left[ \tau D(P_{SL} \parallel Q_{SL}) + E_r(\tau H_P(S), \tau H_P(L|S), W_{YZ|X}) \right] \quad (72)$$

and

$$E_J(Q_{SL}, W_{YZ|X}, \tau) \leq \inf_{P_{SL}} \left[ \tau D(P_{SL} \parallel Q_{SL}) + E(\tau H_P(S), \tau H_P(L|S), W_{YZ|X}) \right] \quad (73)$$

where  $E_r(R_1, R_2, W_{YZ|X})$  is given by  $E_r(R_1, R_2, W_{YZ|UX})$  in (35) with  $|\mathcal{U}| = 1$ , and  $E(R_1, R_2, W_{YZ|X})$  is the channel error exponent for the ABC  $W_{YZ|X}$ .

## 6 Evaluation of the Bounds for $E_J$ : CS over Symmetric AMAC

We established the lower and upper bounds for the JSCC error exponent of the asymmetric 2-user JSCC system. However, we are not able to simplify these bounds for general 2-user JSCC systems (not even for general CS-AMAC and CS-ABC systems) into computable parametric forms as we did for the point-to-point systems [29, 30]. In the following, we only address a special case of CS-AMAC systems where the channel admits a symmetric transition probability distribution. We first introduce the parametric forms of functions  $E_r(R_1, R_2, W_{Y|UX})$  and  $E_{sp}(R_1, R_2, W_{Y|UX})$  defined in (66) and (67), respectively. For any  $R_1, R_2 > 0$ , rewrite

$$E_Y(R_1, R_2, W_{Y|UX}, P_{UX}) = \min \left\{ E_r^{(1)}(R_1 + R_2, W_{Y|UX}, P_{UX}), E_r^{(2)}(R_2, W_{Y|UX}, P_{UX}) \right\}$$

where

$$E_r^{(1)}(R, W_{Y|UX}, P_{UX}) \triangleq \min_{V_{Y|UX}} \left[ D(V_{Y|UX} \parallel W_{Y|UX} | P_{UX}) + \left| I_{P_{UX} V_{Y|UX}}(U, X; Y) - R \right|^+ \right] \quad (74)$$

and

$$E_r^{(2)}(R, W_{Y|UX}, P_{UX}) \triangleq \min_{V_{Y|UX}} \left[ D(V_{Y|UX} \parallel W_{Y|UX} | P_{UX}) + \left| I_{P_{UX} V_{Y|UX}}(X; Y|U) - R \right|^+ \right]. \quad (75)$$

Also, rewrite

$$E_{sp}(R_1, R_2, W_{Y|UX}) = \max_{P_{UX}} E_{sp}(R_1, R_2, W_{Y|UX}, P_{UX})$$

where

$$E_{sp}(R_1, R_2, W_{Y|UX}, P_{UX}) = \min \left\{ E_{sp}^{(1)}(R_1 + R_2, W_{Y|UX}, P_{UX}), E_{sp}^{(2)}(R_2, W_{Y|UX}, P_{UX}) \right\}$$

where

$$E_{sp}^{(1)}(R, W_{Y|UX}, P_{UX}) \triangleq \min_{V_{Y|UX}} \left( D(V_{Y|UX} \parallel W_{Y|UX} | P_{UX}) : I_{P_{UX} V_{Y|UX}}(U, X; Y) \leq R \right) \quad (76)$$

and

$$E_{sp}^{(2)}(R, W_{Y|UX}, P_{UX}) \triangleq \min_{V_{Y|UX}} \left( D(V_{Y|UX} \| W_{Y|UX} | P_{UX}) : I_{P_{UX}V_{Y|UX}}(X; Y|U) \leq R \right). \quad (77)$$

Note that  $E_r^{(1)}$  and  $E_r^{(2)}$  (respectively  $E_{sp}^{(1)}$  and  $E_{sp}^{(2)}$ ) are the random-coding (respectively sphere-packing) type exponents expressed in terms of constrained Kullback-Leibler divergences and mutual informations [10]. In fact, it has been shown in [2] that

$$E_{sp}^{(i)}(R, W_{Y|UX}, P_{UX}) = \max_{\rho \geq 0} [E_i(\rho, W_{Y|UX}, P_{UX}) - \rho R], \quad i = 1, 2,$$

where

$$E_1(\rho_1, W_{Y|UX}, P_{UX}) \triangleq -\log_2 \sum_{y \in \mathcal{Y}} \left( \sum_{(u,x) \in \mathcal{U} \times \mathcal{X}} P_{UX}(u, x) W_{Y|UX}(y|u, x)^{\frac{1}{1+\rho_1}} \right)^{1+\rho_1}, \quad (78)$$

and

$$E_2(\rho_2, W_{Y|UX}, P_{UX}) = -\log_2 \sum_{u \in \mathcal{U}} P_U(u) \sum_{y \in \mathcal{Y}} \left( \sum_{x \in \mathcal{X}} P_{X|U}(x|u) W_{Y|UX}(y|u, x)^{\frac{1}{1+\rho_2}} \right)^{1+\rho_2}. \quad (79)$$

Analogously to [10, Lemma 5.4, Corollary 5.4, p. 168], we can prove the following results; some of them has been proved in [2].

**Lemma 4** Let  $i = 1, 2$ .  $E_r^{(i)}(R, W_{Y|UX}, P_{UX})$  coincides with  $E_{sp}^{(i)}(R, W_{Y|UX}, P_{UX})$  if  $R \geq R_{cr}^{(i)}(W_{Y|UX}, P_{UX})$  where

$$R_{cr}^{(i)}(W_{Y|UX}, P_{UX}) = \left. \frac{\partial E_i(\rho, W_{Y|UX}, P_{UX})}{\partial \rho} \right|_{\rho=1},$$

and is a straight line tangent on  $E_{sp}^{(i)}(R, W_{Y|UX}, P_{UX})$  with slope  $-1$  if  $R \leq R_{cr}^{(i)}(W_{Y|UX}, P_{UX})$ , i.e.

$$E_r^{(i)}(R, W_{Y|UX}, P_{UX}) = \begin{cases} E_{sp}^{(i)}(R, W_{Y|UX}, P_{UX}), & \text{if } R \geq R_{cr}^{(i)}(W_{Y|UX}, P_{UX}), \\ E_{sp}^{(i)}(R_{cr}^{(i)}(W_{Y|UX}, P_{UX}), W_{Y|UX}, P_{UX}) + R_{cr}^{(i)}(W_{Y|UX}, P_{UX}) - R, & \text{if } 0 < R \leq R_{cr}^{(i)}(W_{Y|UX}, P_{UX}). \end{cases}$$

Furthermore,  $E_r^{(i)}(R, W_{Y|UX}, P_{UX})$  has the parametric form

$$E_r^{(i)}(R, W_{Y|UX}, P_{UX}) = \max_{0 \leq \rho \leq 1} [E_i(\rho, W_{Y|UX}, P_{UX}) - \rho R]$$

where  $E_1(\rho, W_{Y|UX}, P_{UX})$  and  $E_2(\rho, W_{Y|UX}, P_{UX})$  are given in (78) and (79) respectively.

Therefore, we can write the functions  $E_r(R_1, R_2, W_{Y|UX})$  in (66) and  $E_{sp}(R_1, R_2, W_{Y|UX})$  in (67) as follows.

$$E_r(R_1, R_2, W_{Y|UX}) = \max_{P_{UX}} \min_{i=1,2} \max_{0 \leq \rho \leq 1} [E_i(\rho, W_{Y|UX}, P_{UX}) - \rho \hat{R}_i] \quad (80)$$

and

$$E_{sp}(R_1, R_2, W_{Y|UX}) = \max_{P_{UX}} \min_{i=1,2} \max_{\rho \geq 0} [E_i(\rho_i, W_{Y|UX}, P_{UX}) - \rho \hat{R}_i] \quad (81)$$

where  $\widehat{R}_1 = R_1 + R_2$  and  $\widehat{R}_2 = R_2$ . Since it is in general hard to find the optimizing solution  $P_{UX}$  for  $E_r$  and  $E_{sp}$  above, we next confine our attention to multiple access channels with some symmetric distributions.

**Definition 3** [2] We say that the multiple access channel  $W_{Y|UX}$  is  $U$ -symmetric if for every  $u \in \mathcal{U}$  the transition matrix  $W_{Y|UX}(\cdot|u, \cdot)$  is symmetric in the sense that the rows (respectively columns) are permutations of each other. An  $X$ -symmetric multiple access channel is defined similarly. We then say that  $W_{Y|UX}$  is symmetric if it is both  $U$ -symmetric and  $X$ -symmetric.

It follows that the multiple access channel with additive noise is symmetric (e.g., see the example below), where a multiple access channel  $W_{Y|UX}$  with (modulo  $B$ ) additive noise  $\{P_F : \mathcal{F}\}$  is described as

$$Y_i = U_i \oplus X_i \oplus F_i \pmod{B}$$

where  $Y_i \in \mathcal{Y}$ ,  $X_i \in \mathcal{X}$ ,  $U_i \in \mathcal{U}$  and  $F_i \in \mathcal{F}$  are the channel's output, two input and noise symbols at time  $i$  such that  $\mathcal{Y} = \mathcal{U} = \mathcal{X} = \mathcal{F} = \{0, 1, 2, \dots, B-1\}$ , and  $F_i$  is independent of  $X_i$  and  $U_i$ ,  $i = 1, 2, \dots, n$ .

It is shown in [2] that if the multiple access channel  $W_{Y|UX}$  is  $U$ -symmetric, then the outer maximum of (80) and (81) is achieved by a joint distribution of the form  $P_{UX}(u, x) = P_U(u)/|\mathcal{X}|$  for every  $x$  and  $u$ . It then follows that for the symmetric multiple access channel, the maximum of (80) and (81) is achieved by a uniform joint distribution

$$P_{UX}^*(u, x) = \frac{1}{|\mathcal{U}||\mathcal{X}|},$$

which is independent of  $\rho$ . Substituting  $P_{UX}^*$  in (80) and (81) yields

$$E_r(R_1, R_2, W_{Y|UX}) = \min_{i=1,2} \max_{0 \leq \rho \leq 1} [\widetilde{E}_i(\rho, W_{Y|UX}) - \rho \widehat{R}_i] \quad (82)$$

and

$$E_{sp}(R_1, R_2, W_{Y|UX}) = \min_{i=1,2} \max_{\rho \geq 0} [\widetilde{E}_i(\rho, W_{Y|UX}) - \rho \widehat{R}_i] \quad (83)$$

where  $\widehat{R}_1 = R_1 + R_2$ ,  $\widehat{R}_2 = R_2$ ,

$$\widetilde{E}_1(\rho, W_{Y|UX}) = (1 + \rho) \log_2(|\mathcal{U}||\mathcal{X}|) - \log_2 \sum_{y \in \mathcal{Y}} \left( \sum_{(u,x) \in \mathcal{U} \times \mathcal{X}} W_{Y|UX}(y|u, x)^{\frac{1}{1+\rho}} \right)^{1+\rho}$$

and

$$\widetilde{E}_2(\rho, W_{Y|UX}) = (1 + \rho) \log_2 |\mathcal{X}| + \log_2 |\mathcal{U}| - \log_2 \sum_{(u,y) \in \mathcal{U} \times \mathcal{Y}} \left( \sum_{x \in \mathcal{X}} W_{Y|UX}(y|u, x)^{\frac{1}{1+\rho}} \right)^{1+\rho}.$$

We also can prove the following identities using a standard optimization method (cf. [29]).

**Lemma 5**

$$\min_{P_{SL}: H_P(S,L)=R} D(P_{SL} \| Q_{SL}) = \max_{\rho \geq 0} [\rho R - E_{s1}(\rho, Q_{SL})], \quad (84)$$

$$\min_{P_{SL}: H_P(L|S)=R} D(P_{SL} \| Q_{SL}) = \max_{\rho \geq 0} [\rho R - E_{s2}(\rho, Q_{SL})], \quad (85)$$

where

$$E_{s1}(\rho, Q_{SL}) = (1 + \rho) \log_2 \sum_{(s,l) \in \mathcal{S} \times \mathcal{L}} Q_{SL}(s, l)^{\frac{1}{1+\rho}}$$

and

$$E_{s2}(\rho, Q_{SL}) = (1 + \rho) \sum_{s \in \mathcal{S}} Q_S(s) \log_2 \sum_{l \in \mathcal{L}} Q_{L|S}(l|s)^{\frac{1}{1+\rho}}.$$

Note that  $E_{s1}(\rho, Q_{SL})$  and  $E_{s2}(\rho, Q_{SL})$  are both concave in  $\rho$ . Clearly, if the marginal distribution  $Q_S(s)$  is uniform, then (84) and (85) are equal. Using (82) we now can write (64) as

$$\begin{aligned} & \min_{P_{SL}} [\tau D(P_{SL} \parallel Q_{SL}) + E_r(\tau H_P(S), \tau H_P(L|S), W_{Y|UX})] \\ = & \min \left\{ \min_{P_{SL}} \left[ \tau D(P_{SL} \parallel Q_{SL}) + \max_{0 \leq \rho_1 \leq 1} [\tilde{E}_1(\rho_1, W_{Y|UX}) - \rho_1 \tau H_P(S, L)] \right], \right. \\ & \left. \min_{P_{SL}} \left[ \tau D(P_{SL} \parallel Q_{SL}) + \max_{0 \leq \rho_2 \leq 1} [\tilde{E}_2(\rho_2, W_{Y|UX}) - \rho_2 \tau H_P(L|S)] \right] \right\} \\ = & \min \left\{ \min_R \left[ \min_{P_{SL}: \tau H_P(S, L) = R} \tau D(P_{SL} \parallel Q_{SL}) + \max_{0 \leq \rho_1 \leq 1} [\tilde{E}_1(\rho_1, W_{Y|UX}) - \rho_1 R] \right], \right. \\ & \left. \min_R \left[ \min_{P_{SL}: \tau H_P(L|S) = R} \tau D(P_{SL} \parallel Q_{SL}) + \max_{0 \leq \rho_2 \leq 1} [\tilde{E}_2(\rho_2, W_{Y|UX}) - \rho_2 R] \right] \right\} \end{aligned} \quad (86)$$

and similarly using (83) we can write (65) as

$$\begin{aligned} & \inf_{P_{SL}} [\tau D(P_{SL} \parallel Q_{SL}) + E_{sp}(\tau H_P(S), \tau H_P(L|S), W_{Y|UX})] \\ = & \min \left\{ \inf_R \left[ \min_{P_{SL}: \tau H_P(S, L) = R} \tau D(P_{SL} \parallel Q_{SL}) + \max_{\rho_1 \geq 0} [\tilde{E}_1(\rho_1, W_{Y|UX}) - \rho_1 R] \right], \right. \\ & \left. \inf_R \left[ \min_{P_{SL}: \tau H_P(L|S) = R} \tau D(P_{SL} \parallel Q_{SL}) + \max_{\rho_2 \geq 0} [\tilde{E}_2(\rho_2, W_{Y|UX}) - \rho_2 R] \right] \right\}. \end{aligned} \quad (87)$$

Consequently, using an optimization technique based on Fenchel duality [29] and (84) and (85), we obtain the following.

**Theorem 5** Given  $Q_{SL}$ , a symmetric  $W_{Y|UX}$ , and the transmission rate  $\tau$ , the lower bound of the JSCC error exponent given in (64) and the upper bound given in (68) can be equivalently expressed as

$$\begin{aligned} \min_{i=1,2} \max_{0 \leq \rho \leq 1} [\tilde{E}_i(\rho, W_{Y|UX}) - \tau E_{si}(\rho, Q_{SL})] & \leq E_J(Q_{SL}, W_{Y|UX}, \tau) \\ & \leq \min_{i=1,2} \max_{\rho \geq 0} [\tilde{E}_i(\rho, W_{Y|UX}) - \tau E_{si}(\rho, Q_{SL})]. \end{aligned} \quad (88)$$

**Example 1** Now consider binary CS  $Q_{SL}$  with distribution

$$\begin{aligned} Q_{SL}(S=0, L=0) &= \frac{2(1-q)}{3}, & Q_{SL}(S=1, L=0) &= \frac{q}{2}, \\ Q_{SL}(S=0, L=1) &= \frac{q}{2}, & Q_{SL}(S=1, L=1) &= \frac{1-q}{3}, \end{aligned}$$

where  $0 < q < 1/2$ . Then

$$E_{s1}(\rho, Q_{SL}) = (1 + \rho) \log_2 \left\{ \left[ \left( \frac{2}{3} \right)^{\frac{1}{1+\rho}} + \left( \frac{1}{3} \right)^{\frac{1}{1+\rho}} \right] (1 - q)^{\frac{1}{1+\rho}} + 2 \left( \frac{q}{2} \right)^{\frac{1}{1+\rho}} \right\},$$

$$E_{s2}(\rho, Q_{SL}) = (1 + \rho) \left( \frac{2(1-q)}{3} + \frac{q}{2} \right) \log_2 \left[ \left( \frac{\frac{2(1-q)}{3}}{\frac{2(1-q)}{3} + \frac{q}{2}} \right)^{\frac{1}{1+\rho}} + \left( \frac{\frac{q}{2}}{\frac{2(1-q)}{3} + \frac{q}{2}} \right)^{\frac{1}{1+\rho}} \right]$$

$$+ (1 + \rho) \left( \frac{1-q}{3} + \frac{q}{2} \right) \log_2 \left[ \left( \frac{\frac{1-q}{3}}{\frac{1-q}{3} + \frac{q}{2}} \right)^{\frac{1}{1+\rho}} + \left( \frac{\frac{q}{2}}{\frac{1-q}{3} + \frac{q}{2}} \right)^{\frac{1}{1+\rho}} \right].$$

Consider a binary multiple access channel  $W_{Y|UX}$  with binary additive noise  $P_F(F = 1) = \epsilon$  ( $0 < \epsilon < 1/2$ ). That is, the transition probabilities are given by

$$\begin{aligned} P_{Y|UX}(Y = 0|U = 0, X = 0) &= 1 - \epsilon, & P_{Y|UX}(Y = 1|U = 0, X = 0) &= \epsilon \\ P_{Y|UX}(Y = 0|U = 0, X = 1) &= \epsilon, & P_{Y|UX}(Y = 1|U = 0, X = 1) &= 1 - \epsilon \\ P_{Y|UX}(Y = 0|U = 1, X = 0) &= \epsilon, & P_{Y|UX}(Y = 1|U = 1, X = 0) &= 1 - \epsilon \\ P_{Y|UX}(Y = 0|U = 1, X = 1) &= 1 - \epsilon, & P_{Y|UX}(Y = 1|U = 1, X = 1) &= \epsilon. \end{aligned}$$

It follows that

$$\tilde{E}_1(\rho, W_{Y|UX}) = \tilde{E}_2(\rho, W_{Y|UX}) = \rho - (1 + \rho) \log_2 \left( \epsilon^{\frac{1}{1+\rho}} + (1 - \epsilon)^{\frac{1}{1+\rho}} \right).$$

In Fig. 6, we plot the lower and upper bounds for the JSCC error exponent  $E_J$  for different  $(q, \epsilon)$  pairs with transmission rate  $t = 0.25$  and  $0.35$ . As illustrated, the upper and lower bounds coincide (this can also be proved by checking that the two outer minimums in (88) are achieved by the same  $i$  and that the inner maximum in the upper bound is achieved by  $\rho \leq 1$ ) for many  $(q, \epsilon)$  pairs (e.g., when  $\tau = 0.25, q = 0.1, \epsilon \geq 0.0205$  and when  $\tau = 0.35, q = 0.1, \epsilon \geq 0.0056$ ), and hence exactly determine the exponent.

## 7 Tandem Coding Error Exponent for the Asymmetric 2-User System

### 7.1 Tandem System with Common Randomization

In Section 4.4, we showed that the reliable transmissibility condition  $(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}(W_{YZ|UX})$  in Theorem 3 can be achieved by a tandem coding system where separately designed source and channel coding operations are sequentially applied; see Figs. 4 and 5 with  $\pi_f$  and  $\pi_g$  being identity mappings. By “separately designed” we mean that the source code is designed without the knowledge of the channel statistics and the channel code is designed without the knowledge of the source statistics. Note however that, as long as the source encoder is directly concatenated by a channel encoder (i.e., if  $\pi_f$  and  $\pi_g$  are

identity mappings), the source statistics would be automatically brought into the channel coding stage. Thus, the performance of the channel code is affected by that of the source code (since the compressed messages (indices) fed into the channel encoders are not necessarily uniformly distributed). To statistically decouple the source and channel coding operations, we need to employ common randomization between the source and channel coding components (e.g., [18]). This results in a “complete” tandem coding system with fully separate source and channel coding operations, and for which we can establish an expression for its error exponent in terms of the source coding and channel coding exponents.

The tandem coding system is depicted in Figs. 4 and 5. As in Section 4.4, the encoder  $f_n$  is composed of two source encoders  $f_{sn}$  and  $g_{sn}$  and one channel encoder  $f_{cn}$ . The difference is that the indices  $i = f_{sn}(\mathbf{1})$  and  $j = g_{sn}(\mathbf{s})$  are separately mapped to channel indices through permutation functions  $\pi_f : \{1, 2, \dots, M_l\} \rightarrow \{1, 2, \dots, M_l\}$  and  $\pi_g : \{1, 2, \dots, M_s\} \rightarrow \{1, 2, \dots, M_s\}$ , which are usually called index assignments ( $\pi_f$  and  $\pi_g$  are assumed to be known at both the transmitter and the receiver). Furthermore, the choice of  $\pi_f$  ( $\pi_g$ , respectively), is assumed random (independent of the source and the channel) and equally likely from all  $M_l!$  ( $M_s!$ , respectively) different possible index assignments, so that the indices fed into the channel encoder have a uniform distribution and are mutually independent:

$$\begin{aligned} \Pr(\pi_f(f_{sn}(L^{\tau n})) = a) &= \sum_{i=1}^{M_l} \Pr(f_{sn}(L^{\tau n}) = i) \Pr(\pi_f(i) = a | f_{sn}(L^{\tau n}) = i) \\ &= \sum_{i=1}^{M_l} \Pr(f_{sn}(L^{\tau n}) = i) \frac{(M_l - 1)!}{M_l!} = \frac{1}{M_l}, \\ \Pr(\pi_g(g_{sn}(S^{\tau n})) = b) &= \frac{1}{M_s}, \\ \Pr(\pi_g(f_{sn}(L^{\tau n})) = a, \pi_g(g_{sn}(S^{\tau n})) = b) &= \Pr(\pi_f(f_{sn}(L^{\tau n})) = a) \Pr(\pi_g(g_{sn}(S^{\tau n})) = b), \end{aligned}$$

for any  $(a, b) \in \{1, 2, \dots, M_l\} \times \{1, 2, \dots, M_s\}$ . Hence common randomization achieves statistical separation between the source and channel coding operations (in the sense that the channel coding error probability is not a function of the source statistics and the source coding error probability is not a function of the channel statistics when no channel decoding error occurs).

Similarly, the encoder  $g_n$  is independently composed of a source encoder  $g_{sn}$ , an index mapping  $\pi_g : \{1, 2, \dots, M_s\} \rightarrow \{1, 2, \dots, M_s\}$ , and a channel encoder  $g_{cn} : \{1, 2, \dots, M_s\} \rightarrow \mathcal{U}^n$ .

At the receiver side, the decoder  $\varphi_n$  is composed of a channel decoder  $\varphi_{cn}$ , a pair of index mappings  $(\pi_f^{-1}, \pi_g^{-1})$  which maps every channel index pair  $(\pi_f(\hat{i}), \pi_g(\hat{j}))$  back to a source index pair  $(\hat{i}, \hat{j})$ , and a source decoder  $\varphi_{sn}$  which outputs the approximation of the source messages  $\mathbf{s}'$  and  $\mathbf{I}'$ . Similarly, the decoder  $\psi_n$  is composed of a channel decoder  $\psi_{cn} : \mathcal{Z}^n \rightarrow \{1, 2, \dots, M_s\}$ , an index mapping  $\pi_g^{-1}$ , and a source decoder  $\psi_{sn} : \{1, 2, \dots, M_s\} \rightarrow \mathcal{S}^{\tau n}$ .

## 7.2 A Formula for the Tandem Coding Error Exponent

We now can study the error performance and exponent of tandem source-channel coding (with common randomization) for the asymmetric 2-user system. Since the tandem code consists of a source code  $(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})$  and a channel code  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$ , we first define the corresponding source coding error exponent (note that the corresponding channel coding error exponent for the asymmetric 2-user channel was defined in Section 4.5).

**Definition 4** The 2-user source coding error exponent  $E(R_1, R_2, Q_{SL})$ , for any  $R_1 > 0$  and  $R_2 > 0$ , is defined by the supremum of the set of all numbers  $E_s$  for which there exists a sequence of source codes  $(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})$  with blocklength  $n$ , common rate no larger than  $R_1$ , and private rate no larger than  $R_2$ , such that

$$E_s \leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{es}^{(n)}(R_1, R_2, Q_{SL}), \quad (89)$$

where  $P_{es}^{(n)}(R_1, R_2, Q_{SL})$  is the source coding probability of error defined in (51).

Denote the probabilities of  $Y$ - and  $Z$ - error for the source coding by

$$P_{Y_{es}}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \triangleq \Pr(\{\varphi_{sn}(i, j) \neq (S^{\tau n}, L^{\tau n})\}) = \sum_{(\mathbf{s}, \mathbf{l}): \psi_{sn}(i, j) \neq (\mathbf{s}, \mathbf{l})} Q_{SL}^{(n)}(\mathbf{s}, \mathbf{l}) \quad (90)$$

and

$$P_{Z_{es}}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) = P_{Z_e}^{(n)}(\widehat{R}_s, Q_S) \triangleq \Pr(\{\psi_{sn}(j) \neq S^{\tau n}\}) = \sum_{\mathbf{s}: \psi_{sn}(i) \neq \mathbf{s}} Q_S^{(n)}(\mathbf{s}) \quad (91)$$

where  $i \triangleq f_{sn}(\mathbf{l})$  and  $j \triangleq g_{sn}(\mathbf{s})$ . Clearly, for any sequence of source codes  $(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})$ , the error probability  $P_{es}^{(n)}(R_1, R_2, Q_{SL})$  must be larger than  $P_{Y_{es}}^{(n)}(R_1, R_2, Q_{SL})$  and  $P_{Z_{es}}^{(n)}(R_1, R_2, Q_{SL})$  but less than the sum of the two; so we have

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{es}^{(n)}(R_1, R_2, Q_{SL}) = \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \max \left( P_{Y_{es}}^{(n)}(R_1, R_2, Q_{SL}), P_{Z_{es}}^{(n)}(R_1, R_2, Q_{SL}) \right). \quad (92)$$

In what follows we need to make three assumptions in order to analyze the probability of error of the overall tandem system. The first two assumptions (referred to as (A1) and (A2)) are regarding the source codes. Let the source codebook for  $(g_{sn}, \psi_{sn})$  (Receiver  $Z$ ) be  $\mathcal{C}^{(g)} = \{\mathbf{c}_1^{(g)}, \dots, \mathbf{c}_{M_l}^{(g)}\} \subseteq \mathcal{S}^{\tau n}$ , and let the source codebook for  $(f_{sn}, g_{sn}, \varphi_{sn})$  (Receiver  $Y$ ) be  $\mathcal{C}^{(f)} \times \mathcal{C}^{(g)}$  where  $\mathcal{C}^{(f)} = \{\mathbf{c}_1^{(f)}, \dots, \mathbf{c}_{M_s}^{(f)}\} \subseteq \mathcal{L}^{\tau n}$ .

- We assume that (A1) the source encoder  $f_{sn}$  satisfies the condition (for every  $n$ ):  $Q_L^{\tau n}(f_{sn}^{-1}(i)) > 0$  and  $\mathbf{c}_i^{(f)} \in f_{sn}^{-1}(i)$  for every  $i = 1, 2, \dots, M_l$ , where  $f_{sn}^{-1}(i) \triangleq \{\mathbf{l} \in \mathcal{L}^{\tau n} : f_{sn}(\mathbf{l}) = i\}$ . Clearly, the assumption has practical meaning. If  $Q_L^{\tau n}(f_{sn}^{-1}(i)) = 0$  for some  $i$ , then the codeword  $\mathbf{c}_i^{(f)}$  is redundant, and we can remove it from the codebook  $\mathcal{C}^{(f)}$ . If  $\mathbf{c}_i^{(f)} \notin f_{sn}^{-1}(i)$ , we can map the index  $i$  to some source message  $\widehat{\mathbf{l}}$  such that  $Q_L^{\tau n}(\widehat{\mathbf{l}}) > 0$  and  $f_{sn}(\widehat{\mathbf{l}}) = i$ , so that the source coding probability of error  $P_{Y_{es}}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL})$  is strictly reduced by setting  $\widehat{\mathbf{l}}$  as the codeword  $\mathbf{c}_i^{(f)}$  (note that  $P_{Z_{es}}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL})$  is independent of  $f_{sn}$ ).



- Similarly, we assume that (A2) the source code  $g_{sn}$  satisfies the condition (for every  $n$ ):  $Q_S^{\tau n}(g_{sn}^{-1}(j)) > 0$  and  $\mathbf{c}_j^{(g)} \in g_{sn}^{-1}(j)$  for every  $j = 1, 2, \dots, M_s$ , where  $g_{sn}^{-1}(j) \triangleq \{\mathbf{s} \in \mathcal{S}^{\tau n} : g_{sn}(\mathbf{s}) = j\}$ . If  $Q_S^{\tau n}(g_{sn}^{-1}(j)) = 0$  for some  $j$ , then the codeword  $\mathbf{c}_j^{(g)}$  is redundant, and we can remove it from the codebook  $\mathcal{C}^{(g)}$ . If  $\mathbf{c}_j^{(g)} \notin g_{sn}^{-1}(j)$ , we can map the index  $j$  to some source message  $\widehat{\mathbf{s}}$  such that  $Q_S^{\tau n}(\widehat{\mathbf{s}}) > 0$  and  $g_{sn}(\widehat{\mathbf{s}}) = j$ , so that the source coding error probabilities  $P_{Y_{es}}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL})$  and  $P_{Z_{es}}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL})$  are strictly reduced by setting  $\widehat{\mathbf{s}}$  as the codeword  $\mathbf{c}_j^{(g)}$ .
- We assume that the limits  $\lim_{n \rightarrow \infty} \frac{1}{n} \log_2 M_l$  and  $\lim_{n \rightarrow \infty} \frac{1}{n} \log_2 M_s$  exist, i.e.,  $\liminf_{n \rightarrow \infty} \frac{1}{n} \log_2 M_l = \limsup_{n \rightarrow \infty} \frac{1}{n} \log_2 M_l$  and  $\liminf_{n \rightarrow \infty} \frac{1}{n} \log_2 M_s = \limsup_{n \rightarrow \infty} \frac{1}{n} \log_2 M_s$ . This assumption is used later to upper bound the tandem coding error exponent in Theorem 6.

We remark that the source code satisfying (A1) and (A2) does not lose optimality in the sense of achieving the source error exponent.

Denote  $\pi^{-1}(i, j) \triangleq (\pi_f^{-1}(i), \pi_g^{-1}(j))$ . By introducing (A1) and (A2), the error probability of the tandem code  $(f_n^*, \varphi_n^*) \triangleq (f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn}, f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$  is given by

$$\begin{aligned}
& P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \\
& \triangleq \Pr \left( \{\varphi_{sn}[\pi^{-1}(\varphi_{cn}(Y^n))] \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_{sn}[\pi_g^{-1}(\psi_{cn}(Z^n))] \neq S^{\tau n}\} \right) \\
& = \sum_{a=1}^{M_l} \sum_{b=1}^{M_s} \underbrace{\Pr(\pi_f[f_{sn}(L^{\tau n})] = a)}_{=1/M_l} \underbrace{\Pr(\pi_g[g_{sn}(S^{\tau n})] = b)}_{=1/M_s} \\
& \quad \left[ \Pr \left( \{\varphi_{cn}(Y^n) \neq (a, b)\} \cup \{\psi_{cn}(Z^n) \neq b\} \mid \pi_f[f_{sn}(L^{\tau n})] = a, \pi_g[g_{sn}(S^{\tau n})] = b \right) + \right. \\
& \quad \left. \Pr \left( \{\varphi_{cn}(Y^n) = (a, b) \text{ and } \psi_{cn}(Z^n) = b\} \cap \{\varphi_{sn}[\pi^{-1}(a, b)] \neq (S^{\tau n}, L^{\tau n}) \text{ or } \psi_{sn}[\pi_g^{-1}(b)] \neq S^{\tau n}\} \mid \right. \right. \\
& \quad \quad \left. \left. \pi_f[f_{sn}(L^{\tau n})] = a, \pi_g[g_{sn}(S^{\tau n})] = b \right) \right] \tag{93}
\end{aligned}$$

$$\begin{aligned}
& = \sum_{a=1}^{M_l} \sum_{b=1}^{M_s} \frac{1}{M_l M_s} \Pr \left( \{\varphi_{cn}(Y^n) \neq (a, b)\} \cup \{\psi_{cn}(Z^n) \neq b\} \mid (a, b) \text{ is sent} \right) \\
& \quad + \Pr \left( \{\varphi_{sn}[S^{\tau n}, L^{\tau n}] \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_{sn}[S^{\tau n}] \neq S^{\tau n}\} \right) \\
& \quad \sum_{a=1}^{M_l} \sum_{b=1}^{M_s} \frac{1}{M_l M_s} \Pr \left( \{\varphi_{cn}(Y^n) = (a, b)\} \cap \{\psi_{cn}(Z^n) = b\} \mid (a, b) \text{ is sent} \right) \tag{94}
\end{aligned}$$

$$= P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) + [1 - P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX})] P_{es}^{(n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}), \tag{95}$$

where (93) follows from assumptions (A1) and (A2), which imply that a channel decoding error must cause an overall system decoding error.

**Definition 5** The tandem coding error exponent  $E_T(Q_{SL}, W_{YZ|UX}, \tau)$  for source  $Q_{SL}$  and channel  $W_{YZ|UX}$  is defined as the supremum of the set of all numbers  $\widehat{E}$  for which there exists a sequence of tandem codes

$(f_n^*, \varphi_n^*)$  satisfying (A1) and (A2) with transmission rate  $\tau$  such that

$$\hat{E} \leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau).$$

When there is no possibility of confusion,  $E_T(Q_{SL}, W_{YZ|UX}, \tau)$  will often be written as  $E_T$ . The following lemma illustrates the relation between  $E_T$  and  $E_J$ .

**Lemma 6**  $E_J(Q_{SL}, W_{YZ|UX}, \tau) \geq E_T(Q_{SL}, W_{YZ|UX}, \tau)$ .

**Proof:** By definition, for any sequence of rate  $\tau$  tandem codes  $\{(f_n^*, \varphi_n^*)\}_{n=1}^\infty$  composed of a sequence of source codes  $\{(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})\}_{n=1}^\infty$  and a sequence of channel codes  $\{(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})\}_{n=1}^\infty$ , we have

$$\begin{aligned} & P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \\ &= \mathbb{E}_{\pi_f, \pi_g} \Pr \left( \{\varphi_{sn} [\pi^{-1}(\varphi_{cn}(Y^n))] \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_{sn} [\pi_g^{-1}(\psi_{cn}(Z^n))] \neq S^{\tau n}\} \mid \pi_f \text{ and } \pi_g \text{ are fixed} \right) \\ &\geq \min_{\pi_f, \pi_g} \Pr \left( \{\varphi_{sn} [\pi^{-1}(\varphi_{cn}(Y^n))] \neq (S^{\tau n}, L^{\tau n})\} \cup \{\psi_{sn} [\pi_g^{-1}(\psi_{cn}(Z^n))] \neq S^{\tau n}\} \mid \pi_f \text{ and } \pi_g \text{ are fixed} \right). \end{aligned}$$

Let the above minimum be achieved by  $\pi_f^* = \pi_f^*(n)$  and  $\pi_g^* = \pi_g^*(n)$ . Obviously, there exists a sequence of JSC codes  $\{(f_n, g_n, \varphi_n, \psi_n)\}_{n=1}^\infty$  where  $f_n$  is composed of  $f_{sn}$ ,  $g_{sn}$ ,  $\pi_f^*$ ,  $\pi_g^*$ ,  $f_{cn}$ ,  $g_{cn}$  is composed of  $g_{sn}$ ,  $\pi_g^*$  and  $g_{cn}$ ,  $\varphi_n$  is composed of  $\varphi_{cn}$ ,  $(\pi_f^{*-1}, \pi_g^{*-1})$ , and  $\varphi_{sn}$ , and finally,  $\psi_n$  is composed of  $\psi_{cn}$ ,  $\pi_g^{*-1}$ , and  $\psi_{sn}$  (cf. Figs. 4 and 5), such that

$$P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \geq P_e^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \quad \text{for any } n \geq 1,$$

where  $P_e^{(n)}(Q_{SL}, W_{YZ|UX}, \tau)$  is the probability of error induced by the JSC codes  $\{(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})\}$ . Since this holds for any sequence of tandem codes (satisfying (A1) and (A2)), it then follows from the definition of joint and tandem exponents that  $E_J \geq E_T$ .  $\blacksquare$

We next derive a formula for  $E_T$  in terms of the corresponding source and channel error exponents.

**Theorem 6**

$$E_T(Q_{SL}, W_{YZ|UX}, \tau) = \sup_{R_1 > 0, R_2 > 0} \min \left\{ \tau e \left( \frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL} \right), E(R_1, R_2, W_{YZ|UX}) \right\}$$

where  $e(R_1, R_2, Q_{SL})$  is the 2-user source coding error exponent defined in (4) and  $E(R_1, R_2, W_{YZ|UX})$  is the asymmetric 2-user channel coding error exponent defined in (2).

**Remark 1** As can be seen from the proof below, the common randomization set-up together with the assumptions regarding the source and channel codes are essentially needed to prove the converse part of the tandem coding error exponent; the forward part (the proof of the lower bound on the exponent) is still valid for tandem systems without these assumptions.

**Proof:**

*Forward Part:* we show that there exists a sequence of tandem codes  $(f_n^*, \varphi_n^*)$  satisfying (A1) and (A2) such that

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) > \sup_{R_1 > 0, R_2 > 0} \min \left\{ \tau e \left( \frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL} \right), E(R_1, R_2, W_{YZ|UX}) \right\} - \delta$$

for any  $\delta > 0$ . It follows from (95) that

$$P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \leq 2 \max \{ P_{es}^{(\tau n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}), P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) \}$$

or equivalently

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \geq \min \left\{ \liminf_{n \rightarrow \infty} -\frac{1}{n} P_{es}^{(\tau n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \right. \\ \left. \liminf_{n \rightarrow \infty} -\frac{1}{n} P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) \right\} \quad (96)$$

Fix  $\delta > 0$  and let  $R_1 = \tau \lim_{n \rightarrow \infty} \widehat{R}_s$  and  $R_2 = \tau \lim_{n \rightarrow \infty} \widehat{R}_l$ . According to the definition of the 2-user source coding error exponent, there exists a sequence of source codes  $(\tilde{f}_{sn}, \tilde{g}_{sn}, \tilde{\varphi}_{sn}, \tilde{\psi}_{sn})$  satisfying (A1) and (A2) (since (A1) and (A2) do not lose optimality) with common source rate  $\widehat{R}_s$  and private source rate  $\widehat{R}_l$  such that

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} P_{es}^{(\tau n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \geq e \left( \frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL} \right) - \delta.$$

On the other hand, according to the definition of the asymmetric 2-user channel coding error exponent, there exists a sequence of channel codes  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$  with common rate  $\tau \widehat{R}_s$  and private rate  $\tau \widehat{R}_l$  such that

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) \geq E(\tau R_1, \tau R_2, W_{YZ|UX}) - \delta.$$

Finally, since the sequences of rates  $\widehat{R}_s$  and  $\widehat{R}_l$  can be arbitrarily choose, and so are  $R_1$  and  $R_2$ , we can take the supremum of  $R_1$  and  $R_2$ , completing the proof of the forward part.

*Converse Part:* We show that for any sequence of tandem codes  $(f_n^*, \varphi_n^*)$  with rate  $\tau$  composed by source codes  $\{(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})\}_{n=1}^{\infty}$  satisfying assumptions (A1) and (A2) and channel codes  $\{(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})\}_{n=1}^{\infty}$ ,

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \leq \sup_{R_1 > 0, R_2 > 0} \min \left\{ \tau e \left( \frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL} \right), E(R_1, R_2, W_{YZ|UX}) \right\}. \quad (97)$$

Let the private index set for the tandem system be  $\{1, 2, \dots, M_l\}$  (cf. Figs. 4 and 5). Thus the private source and channel code rates are given by  $\widehat{R}_l = \frac{1}{\tau n} \log_2 M_l$  and  $R_l = \tau \widehat{R}_l$ , respectively. Let the common index set be  $\{1, 2, \dots, M_s\}$ . Thus the common source code rate and channel code rate are given by  $\widehat{R}_s = \frac{1}{\tau n} \log_2 M_s$  and  $R_s = \tau \widehat{R}_s$ , respectively.

We first assume that  $\limsup_{n \rightarrow \infty} -\frac{1}{n} \log_2 [1 - P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX})] \geq \delta$  for some positive  $\delta$  independent of  $n$ , which implies that there exists a sequence  $n_0 \leq n_1 \leq n_2 \leq \dots \leq \infty$  such that

$$\lim_{i \rightarrow \infty} P_{ec}^{(n_i)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) \geq 1 - \lim_{i \rightarrow \infty} 2^{-n_i \delta} = 1$$

In this trivial case,

$$\begin{aligned} \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) &\leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) \\ &\leq \lim_{i \rightarrow \infty} -\frac{1}{n_i} \log_2 P_{ec}^{(n_i)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}) \\ &= 0 \end{aligned}$$

and (97) holds. Next we assume that  $\limsup_{n \rightarrow \infty} -\frac{1}{n} \log_2 [1 - P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX})] = 0$ . It then follows from (95) that

$$\begin{aligned} \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) &\leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \left[ (1 - P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX})) P_{es}^{(\tau n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \right] \\ &\leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{es}^{(\tau n)}(\widehat{R}_s, \widehat{R}_l, Q_{SL}) \end{aligned} \quad (98)$$

and

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \leq \liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{ec}^{(n)}(\tau \widehat{R}_s, \tau \widehat{R}_l, W_{YZ|UX}). \quad (99)$$

Let

$$R_1 = \lim_{n \rightarrow \infty} \tau \widehat{R}_s = \lim_{n \rightarrow \infty} \frac{\log_2 M_s}{n} \quad (100)$$

and

$$R_2 = \lim_{n \rightarrow \infty} \tau \widehat{R}_l = \lim_{n \rightarrow \infty} \frac{\log_2 M_l}{n}. \quad (101)$$

By definition, the source error exponent  $e\left(\frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL}\right)$  is the largest (supremum) number  $E_s$  such that there exists a sequence of source codes  $(\tilde{f}_{sn}, \tilde{g}_{sn}, \tilde{\varphi}_{sn}, \tilde{\psi}_{sn})$  with message sets  $\{1, 2, \dots, \widetilde{M}_s\}$  and  $\{1, 2, \dots, \widetilde{M}_l\}$  satisfying

$$\begin{aligned} \limsup_{n \rightarrow \infty} \frac{\log_2 \widetilde{M}_s}{\tau n} &\leq \frac{R_1}{\tau}, \\ \limsup_{n \rightarrow \infty} \frac{\log_2 \widetilde{M}_l}{\tau n} &\leq \frac{R_2}{\tau}, \end{aligned}$$

and

$$\liminf_{n \rightarrow \infty} -\frac{1}{\tau n} \log_2 \Pr \left( \{\tilde{\varphi}_{sn}(\tilde{f}_{sn}(L^{\tau n}), \tilde{g}_{sn}(S^{\tau n})) \neq (S^{\tau n}, L^{\tau n})\} \cup \{\tilde{\psi}_{sn}(\tilde{g}_{sn}(S^{\tau n})) \neq S^{\tau n}\} \right) \geq E_s.$$

This means that

$$\liminf_{n \rightarrow \infty} -\frac{1}{\tau n} \log_2 \Pr \left( \{\tilde{\varphi}_{sn}(\tilde{f}_{sn}(L^{\tau n}), \tilde{g}_{sn}(S^{\tau n})) \neq (S^{\tau n}, L^{\tau n})\} \cup \{\tilde{\psi}_{sn}(\tilde{g}_{sn}(S^{\tau n})) \neq S^{\tau n}\} \right) \leq e\left(\frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL}\right)$$

holds for all source codes  $(\tilde{f}_{sn}, \tilde{g}_{sn}, \tilde{\varphi}_{sn}, \tilde{\psi}_{sn})$  with  $\limsup_{n \rightarrow \infty} \frac{\log_2 \widetilde{M}_s}{\tau n} \leq \frac{R_1}{\tau}$  and  $\limsup_{n \rightarrow \infty} \frac{\log_2 \widetilde{M}_l}{\tau n} \leq \frac{R_2}{\tau}$ , and hence holds for the sequence of block codes  $(f_{sn}, g_{sn}, \varphi_{sn}, \psi_{sn})$  with rates  $(\widehat{R}_s, \widehat{R}_l)$  satisfying (A1) and (A2).

Similarly, by the definition of the asymmetric 2-user channel error exponent and on account of (100) and (101) we have

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 \Pr \left( \{\widehat{\varphi}_{cn}(Y^n) \neq (J, I)\} \cup \{\widehat{\psi}_{cn}(Z^n) \neq J\} \right) \leq E(R_1, R_2, W_{YZ|UX})$$

holds for all channel codes  $(\widehat{f}_{cn}, \widehat{g}_{cn}, \widehat{\varphi}_{cn}, \widehat{\psi}_{cn})$  with common and private message sets  $\{1, 2, \dots, \widehat{M}_s\}$  and  $\{1, 2, \dots, \widehat{M}_l\}$  such that  $\liminf_{n \rightarrow \infty} \frac{\log_2 \widehat{M}_s}{n} \geq R_1$  and  $\liminf_{n \rightarrow \infty} \frac{\log_2 \widehat{M}_l}{n} \geq R_2$ , and of course holds for the sequence of channel codes  $(f_{cn}, g_{cn}, \varphi_{cn}, \psi_{cn})$  with common rate  $R_s = \tau \widehat{R}_s$  and private rate  $R_l = \tau \widehat{R}_l$ .

Putting things together, (98) and (99) yield

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log_2 P_{e^*}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \leq \min \left\{ \tau e \left( \frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL} \right), E(R_1, R_2, W_{YZ|UX}) \right\},$$

holds for all the source codes satisfying (A1) and (A2) and all the channel codes with  $\lim_{n \rightarrow \infty} \frac{\log_2 M_s}{n} = R_1$  and  $\lim_{n \rightarrow \infty} \frac{\log_2 M_l}{n} = R_2$ . Since the above is satisfied for any sequences of  $M_s > 0$  and  $M_l > 0$ , and hence for all  $R_1 > 0$  and  $R_2 > 0$ , we take the supremum over  $R_1 > 0, R_2 > 0$  and obtain (97).  $\blacksquare$

### 7.3 Comparison of Joint and Tandem Coding Error Exponents

Although tandem source-channel coding can achieve reliable transmissibility, it might not achieve the system JSCC error exponent. In the following we consider the tandem system consisting of CS  $Q_{SL}$  and AMAC  $W_{Y|UX}$ . For the CS-AMAC tandem system, we have only one receiver, Receiver  $Y$ , and the source decoder (cf. Fig. 5)  $\varphi_{sn}$  becomes a Slepian-Wolf decoder [6]. Furthermore,

$$P_{es}^{(n)}(R_1, R_2, Q_{SL}) = P_{Y_{es}}^{(n)}(R_1, R_2, Q_{SL}) = \sum_{(\mathbf{s}, \mathbf{l}) : \psi_{sn}(i, j) \neq (\mathbf{s}, \mathbf{l})} Q_{SL}^{(n)}(\mathbf{s}, \mathbf{l})$$

and

$$P_{ec}^{(n)}(R_1, R_2, W_{Y|UX}) = P_{Y_{ec}}^{(n)}(R_1, R_2, W_{Y|UX}) = \frac{1}{2^{R_1 + R_2}} \sum_{\mathcal{M}_s \times \mathcal{M}_l} \sum_{\mathbf{y} : \varphi_{cn}(\mathbf{y}) \neq (j, i)} W_{Y|X}^{(n)}(\mathbf{y} | \mathbf{u}, \mathbf{x}).$$

In this case, we can upper bound the source error exponent by

$$e \left( \frac{R_1}{\tau}, \frac{R_2}{\tau}, Q_{SL} \right) \leq \min_{P_{SL} : \tau H_P(S, L) = R_1 + R_2} D(P_{SL} \| Q_{SL}) = \max_{\rho \geq 0} \left[ \rho \frac{R_1 + R_2}{\tau} - E_{s1}(\rho, Q_{SL}) \right], \quad (102)$$

which is obtained by viewing the two source encoders  $f_{sn}$  and  $g_{sn}$  as a joint encoder [11], where  $E_{s1}(\rho, Q_{SL})$  is given by Lemma 5. Therefore, we can upper bound the tandem coding error exponent for the CS-AMAC system by

$$E_T(Q_{SL}, W_{YZ|UX}, \tau) \leq \sup_{R_1 > 0, R_2 > 0} \min \left\{ \max_{\rho \geq 0} [\rho(R_1 + R_2) - \tau E_{s1}(\rho, Q_{SL})], E_{sp}(R_1, R_2, W_{Y|UX}) \right\} \quad (103)$$

where  $E_{sp}(R_1, R_2, W_{Y|UX})$  is an upper bound for the channel error exponent and is given by (81).

**Example 2** Now consider the same binary CS  $Q_{SL}$  given in Example 1 such that

$$E_{s1}(\rho, Q_{SL}) = (1 + \rho) \log_2 \left\{ \left[ \left( \frac{2}{3} \right)^{\frac{1}{1+\rho}} + \left( \frac{1}{3} \right)^{\frac{1}{1+\rho}} \right] (1 - q)^{\frac{1}{1+\rho}} + 2 \left( \frac{q}{2} \right)^{\frac{1}{1+\rho}} \right\},$$

and consider the same binary multiple access channel  $W_{Y|UX}$  as in Example 1 with binary additive noise  $P_F(F = 1) = \epsilon$  ( $0 < \epsilon < 1/2$ ) such that

$$E_{sp}(R_1, R_2, W_{Y|UX}) = \min_{i=1,2} \max_{\rho \geq 0} [\tilde{E}_i(\rho, W_{Y|UX}) - \rho \hat{R}_i] = \max_{\rho \geq 0} [\tilde{E}_i(\rho, W_{Y|UX}) - \rho(R_1 + R_2)]$$

where  $\hat{R}_1 = R_1 + R_2$ ,  $\hat{R}_2 = R_2$ , and

$$\tilde{E}_1(\rho, W_{Y|UX}) = \tilde{E}_2(\rho, W_{Y|UX}) = \rho - (1 + \rho) \log_2 \left( \epsilon^{\frac{1}{1+\rho}} + (1 - \epsilon)^{\frac{1}{1+\rho}} \right).$$

It follows from (103) that the upper bound for  $E_T$  only depends on the sum rate  $R_1 + R_2$  and hence the upper bound can be reduced to

$$E_T(Q_{SL}, W_{YZ|UX}, \tau) \leq \sup_{R > 0} \min \left\{ \max_{\rho \geq 0} [\rho R - \tau E_{s1}(\rho, Q_{SL})], \max_{\rho \geq 0} [\tilde{E}_1(\rho, W_{Y|UX}) - \rho R] \right\}.$$

In Fig. 7, we plot the lower bound for  $E_J$  from (88), and the above upper bound for  $E_T$  for different source and channel parameters. It is seen that for a large class of  $(q, \epsilon)$  pairs with the same transmission rate  $\tau$ , there is a considerable gap between the upper bound for  $E_T$  and the lower bound for  $E_J$ , which implies that JSCC can substantially outperform tandem coding in terms of error exponent for many binary CS-AMAC systems with additive noise. In fact, from Fig. 7, we see that  $E_J$  almost doubles  $E_T$  for many  $(q, \epsilon)$  pairs. When  $E_J \approx 2E_T$  holds, it can be equivalently interpreted that, to achieve the same system error performance, JSCC only requires around half delay of the tandem coding, provided that the coding length is sufficiently large.

## 8 Conclusion

In this paper, we study the error performance and exponents of JSCC for a class of discrete memoryless communication systems which transmit two correlated sources over a 2-transmitter 2-receiver channel in an “asymmetric” way. For such systems, we derive universally achievable error exponent pairs for the two receivers by employing a generalized type-packing lemma. We also establish a lower and an upper bound for the system JSCC error exponent. We next specialize these results to CS-AMAC and CS-ABC systems. As a special case, we study the analytical computation of the lower and upper bounds for CS-AMAC systems for which the channel admits a symmetric conditional distribution. We show that the lower and upper bounds coincide for many binary CS-AMAC source-channel pairs with additive noise, and hence exactly determine the JSCC error exponent.

As a consequence of our lower bound for the JSCC error exponent, we prove a JSCC theorem for the asymmetric 2-user system, i.e., a sufficient and necessary condition for the reliable transmissibility of the two CS over the asymmetric channel is provided. It is demonstrated that the condition can actually be achieved by a tandem coding scheme, which combines separate source and channel coding. This means that tandem coding does not lose optimality from the point of view of reliable transmissibility. Nevertheless, tandem coding might not be optimal in terms of the error exponent. To exploit the advantage of JSCC over tandem coding for the 2-user system, we show that the tandem coding exponent can never be larger than the JSCC exponent and we derive a formula for the tandem exponent in terms of the source and channel coding exponents. The formula holds under two basic assumptions on the source code and the assumption that common randomization is used at the transmitter and receiver sides to render the source and channel coding operations statistically decoupled from one another. By numerically comparing the upper bound for the tandem exponent and the lower bound for the JSCC exponent, we note that there is a considerable gain of the JSCC error exponent over the tandem coding error exponent for a large class of binary CS-AMAC systems with additive noise. Note that this prospective benefit of JSCC over tandem coding can also translate into substantial reductions in system complexity and coding delay.

## A Proof of Lemma 3

Although the result (5) of Lemma 3 was already shown in [8], we include its proof here since we need to show that (5) holds simultaneously with (6) and (7). We employ a random selection argument as used in [8]. For each  $i = 1, 2, \dots, m_n$ , we randomly generate a set of  $2N_i$  sequences (according to a uniform distribution) from the type class  $\mathbb{T}_{A_i} = \mathbb{T}_{P_{A_i}}$ ,  $\mathcal{C}_i \triangleq \{\mathbf{a}_1^{(i)}, \mathbf{a}_2^{(i)}, \dots, \mathbf{a}_{2N_i}^{(i)}\} \subseteq \mathbb{T}_{A_i}$ , i.e., each  $\mathbf{a}_p^{(i)}$  is randomly drawn from the type class  $\mathbb{T}_{A_i}$  with probability  $1/|\mathbb{T}_{A_i}|$ ,  $p = 1, 2, \dots, 2N_i$ . Each set has  $2N_i$  elements rather than  $N_i$  because an expurgation operation will be performed later. Also, we denote the set  $\mathcal{C}_i^p \triangleq \mathcal{C}_i / \{\mathbf{a}_p^{(i)}\}$ .

Now for each  $i$  with associated  $j = j(i) = 1, 2, \dots, m'_{in}$ , we randomly generate  $4N_i M_{ij}$  sequences (according to a uniform distribution)

$$\left\{ \mathbf{b}_{11}^{(j)}, \mathbf{b}_{12}^{(j)}, \dots, \mathbf{b}_{1,2M_{ij}}^{(j)}, \mathbf{b}_{21}^{(j)}, \mathbf{b}_{22}^{(j)}, \dots, \mathbf{b}_{2,2M_{ij}}^{(j)}, \dots, \mathbf{b}_{2N_i,1}^{(j)}, \mathbf{b}_{2N_i,2}^{(j)}, \dots, \mathbf{b}_{2N_i,2M_{ij}}^{(j)} \right\}$$

such that the set

$$\begin{aligned} \mathcal{C}_{ij} \triangleq & \left\{ \left( \mathbf{a}_1^{(i)}, \mathbf{b}_{11}^{(j)} \right), \left( \mathbf{a}_1^{(i)}, \mathbf{b}_{12}^{(j)} \right), \dots, \left( \mathbf{a}_1^{(i)}, \mathbf{b}_{1,2M_{ij}}^{(j)} \right), \right. \\ & \left( \mathbf{a}_2^{(i)}, \mathbf{b}_{21}^{(j)} \right), \left( \mathbf{a}_2^{(i)}, \mathbf{b}_{22}^{(j)} \right), \dots, \left( \mathbf{a}_2^{(i)}, \mathbf{b}_{2,2M_{ij}}^{(j)} \right), \\ & \dots \dots \\ & \left. \left( \mathbf{a}_{2N_i}^{(i)}, \mathbf{b}_{2N_i,1}^{(j)} \right), \left( \mathbf{a}_{2N_i}^{(i)}, \mathbf{b}_{2N_i,2}^{(j)} \right), \dots, \left( \mathbf{a}_{2N_i}^{(i)}, \mathbf{b}_{2N_i,2M_{ij}}^{(j)} \right) \right\} \subseteq \mathbb{T}_{A_i B_j} = \mathbb{T}_{P_{A_i} P_{B_j | A_i}}. \end{aligned}$$

In other words, each  $\mathbf{b}_{p,q}^{(j)}$  is drawn from  $\mathbb{T}_{B_j | A_i}(\mathbf{a}_p^{(i)})$  with probability  $1/|\mathbb{T}_{B_j | A_i}(\mathbf{a}_p^{(i)})|$ ,  $q = 1, 2, \dots, M_{ij}$ , and hence each pair  $(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)})$  is drawn from  $\mathbb{T}_{A_i B_j}$  with probability  $1/|\mathbb{T}_{A_i B_j}|$ . Furthermore, we denote

the set  $\mathcal{C}_{ij}^{pq} \triangleq \mathcal{C}_{ij} / \left\{ \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \right\}$ . For any  $1 \leq i, k \leq m_n$ ,  $1 \leq j \leq m'_{in}$  and  $1 \leq l \leq m'_{kn}$ , define

$$\mathcal{V}_{i,k} \triangleq \left\{ V_{A'|A} \in \mathcal{P}_n(\mathcal{A}|P_{A_i}) : \sum_{a \in \mathcal{A}} P_{A_i}(a) V_{A'|A}(a'|a) = P_{A_k}(a') \right\}$$

and

$$\mathcal{V}_{i,j,k,l} \triangleq \left\{ V_{A'B'|AB} \in \mathcal{P}_n(\mathcal{A} \times \mathcal{B}|P_{A_i B_j}) : \sum_{(a,b) \in \mathcal{A} \times \mathcal{B}} P_{A_i B_j}(a,b) V_{A'B'|AB}(a',b'|a,b) = P_{A_k B_l}(a',b') \right\}.$$

Based on the above set-up, the following inequalities hold.

i.) For any  $(i,j) \neq (k,l)$  and any  $V_{A'B'|AB} \in \mathcal{V}_{i,j,k,l}$ ,

$$\begin{aligned} & \mathbb{E} \left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{kl} \right| \\ & \leq \mathbb{E} \left| \left\{ (p', q') : \left( \mathbf{a}_{p'}^{(k)}, \mathbf{b}_{p',q'}^{(l)} \right) \in \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \right\} \right| \\ & = 4N_k M_{kl} \Pr \left\{ \left( \mathbf{a}_1^{(k)}, \mathbf{b}_{1,1}^{(l)} \right) \in \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \right\} \\ & = 4N_k M_{kl} \frac{\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \right|}{\left| \mathbb{T}_{A_k B_l} \right|} \\ & \leq 4N_k M_{kl} (n+1)^{|\mathcal{A}||\mathcal{B}|} 2^{-nI_{P_{A_i B_j} V_{A'B'|AB}}(A', B'; A, B)}, \end{aligned} \quad (104)$$

where the above expectation and probability are taken over the uniform distribution

$$\widehat{P}_{k,l} \left( \mathbf{a}_{p'}^{(k)}, \mathbf{b}_{p',q'}^{(l)} \right) \triangleq \frac{1}{\left| \mathbb{T}_{A_k B_l} \right|} \quad \forall \quad 1 \leq k \leq m_n, \quad 1 \leq l \leq m'_{kn}, \quad 1 \leq p' \leq N_k, \quad 1 \leq q' \leq M_{kl}, \quad (105)$$

and (104) follows from the basic facts (Lemma 1) that

$$\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \right| \leq 2^{nH_{P_{A_i B_j} V_{A'B'|AB}}(A', B'; A, B)}$$

and that

$$\left| \mathbb{T}_{A_k B_l} \right| \geq (n+1)^{-|\mathcal{A}||\mathcal{B}|} 2^{nH_{P_{A_k B_l}}(A', B')},$$

noting that the marginal distribution of  $P_{A_i B_j} V_{A'B'|AB}$  for RV's  $(A', B')$  is  $P_{A_k B_l}$ .

ii.) For any  $(i,j) = (k,l)$  and any  $V_{A'B'|AB} \in \mathcal{V}_{i,j,i,j}$ , likewise,

$$\mathbb{E} \left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{ij}^{pq} \right| \leq 4N_i M_{ij} (n+1)^{|\mathcal{A}||\mathcal{B}|} 2^{-nI_{P_{A_i B_j} V_{A'B'|AB}}(A', B'; A, B)}, \quad (106)$$

where the expectation is taken over the uniform distribution  $\widehat{P}_{i,j}$  defined by (105).

iii.) For any  $i$  and  $j \neq l$ , and any  $V_{AB'|AB} \in \mathcal{V}_{i,j,il}$ , similarly we have

$$\mathbb{E} \left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{il} \right| \leq 4N_i M_{il} (n+1)^{|\mathcal{A}||\mathcal{B}|} 2^{-nI_{P_{A_i B_j} V_{AB'|AB}}(A, B'; A, B)}.$$



Using the identity

$$I_{P_{A_i B_j} V_{AB'|AB}}(A, B'; A, B) = H_{P_{A_i}}(A) + I_{P_{A_i B_j} V_{AB'|AB}}(B'; B|A)$$

and assumption (3)

$$\frac{1}{n} \log_2 N_i < H_{P_{A_i}}(A) - \delta,$$

we obtain another bound

$$\mathbb{E} \left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{il} \right| \leq 4M_{il}(n+1)^{|\mathcal{A}||\mathcal{B}|} 2^{-nI_{P_{A_i B_j} V_{AB'|AB}}(B'; B|A)}, \quad (107)$$

where the expectation is taken over the uniform distribution  $\widehat{P}_{i,l}$ .

iv.) For any  $i$  and  $j = l$ , and any  $V_{A'B'|AB} \in \mathcal{V}_{ij,il}$ , likewise,

$$\mathbb{E} \left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{ij}^{pq} \right| \leq 4M_{ij}(n+1)^{|\mathcal{A}||\mathcal{B}|} 2^{-nI_{P_{A_i B_j} V_{A'B'|AB}}(B'; B|A)}, \quad (108)$$

where the expectation is taken over the uniform distribution  $\widehat{P}_{i,j}$ .

v.) For any  $i \neq k$  and any  $V_{A'|A} \in \mathcal{V}_{i,k}$ ,

$$\begin{aligned} \mathbb{E} \left| \mathbb{T}_{V_{A'|A}} \left( \mathbf{a}_p^{(i)} \right) \cap \mathcal{C}_k \right| &\leq \mathbb{E} \left| \left\{ p' : \mathbf{a}_{p'}^{(k)} \in \mathbb{T}_{V_{A'|A}} \left( \mathbf{a}_p^{(i)} \right) \right\} \right| \\ &= 2N_k \Pr \left\{ \mathbf{a}_1^{(i)} \in \mathbb{T}_{V_{A'|A}} \left( \mathbf{a}_p^{(i)} \right) \right\} \\ &= 2N_k \frac{\left| \mathbb{T}_{V_{A'|A}} \left( \mathbf{a}_p^{(i)} \right) \right|}{|\mathbb{T}_{A_k}|} \\ &\leq 2N_k (n+1)^{-|\mathcal{A}|} 2^{-nI_{P_{A_i} V_{A'|A}}(A'; A)}, \end{aligned} \quad (109)$$

where the above expectation and probability are taken over the uniform distribution

$$\widetilde{P}_k(\mathbf{a}_{p'}^{(k)}) \triangleq \frac{1}{|\mathbb{T}_{A_k}|}, \quad \forall \quad 1 \leq k \leq m_n, \quad 1 \leq p' \leq N_k, \quad (110)$$

and (109) follows from the basic facts (Lemma 1) that

$$\left| \mathbb{T}_{V_{A'|A}} \left( \mathbf{a}_1^{(i)} \right) \right| \leq 2^{nH_{P_{A_i} V_{A'|A}}(A'|A)}$$

and that

$$|\mathbb{T}_{A_k}| \geq (n+1)^{|\mathcal{A}|} 2^{nH_{P_{A_k}}(A')},$$

noting that the marginal distribution of  $P_{A_i} V_{A'|A}$  for the RV  $A'$  is  $P_{A_k}$ .

vi.) For any  $i = k$  and any  $V_{A'|A} \in \mathcal{V}_{i,k}$ , likewise,

$$\mathbb{E} \left| \mathbb{T}_{V_{A'|A}} \left( \mathbf{a}_p^{(i)} \right) \cap \mathcal{C}_i^p \right| \leq 2N_k (n+1)^{-|\mathcal{A}|} 2^{-nI_{P_{A_i} V_{A'|A}}(A'; A)}, \quad (111)$$

where the expectation is taken over the uniform distribution  $\widetilde{P}_i$  defined in (110).

Note also if  $V_{A'B'|AB} \notin \mathcal{V}_{ij,kl}$

$$\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{kl} \right| = 0,$$

and if  $V_{A'B'|AB} \notin \mathcal{V}_{ij,ij}$

$$\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{ij}^{pq} \right| = 0.$$

Therefore, it follows from (104) and (106) that for any  $V_{A'B'|AB} \in \mathcal{P}_n(\mathcal{A} \times \mathcal{B} | \mathcal{A} \times \mathcal{B})$ ,

$$\begin{aligned} & \frac{\mathbb{E} \left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{x}_{p,q}^{(j)} \right) \cap \mathcal{C}_{ij}^{pq} \right|}{4N_i M_{ij}} + \sum_{(k,l) \neq (i,j)} \frac{\mathbb{E} \left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{kl} \right|}{4N_k M_{kl}} \\ & \leq m_n (\max_i m'_{in}) (n+1) |\mathcal{A}| |\mathcal{B}| 2^{-n I_{P_{A_i B_j} V_{A'B'|AB}}(A', B'; A, B)}. \end{aligned} \quad (112)$$

Taking the sum over all  $V_{A'B'|AB} \in \mathcal{P}_n(\mathcal{A} \times \mathcal{B} | \mathcal{A} \times \mathcal{B})$ , and using the fact (Lemma 1)

$$|\mathcal{P}_n(\mathcal{A} \times \mathcal{B} | \mathcal{A} \times \mathcal{B})| \leq (n+1) |\mathcal{A}|^2 |\mathcal{B}|^2$$

and  $|\mathcal{A}|^2 |\mathcal{B}|^2 + |\mathcal{A}| |\mathcal{B}| \leq 2 |\mathcal{A}|^2 |\mathcal{B}|^2$ , we obtain

$$\mathbb{E} S_{ij}^{pq} \leq (n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n (\max_i m'_{in})$$

where

$$\begin{aligned} S_{ij}^{pq} & \triangleq \sum_{V_{A'B'|AB} \in \mathcal{P}_n(\mathcal{A} \times \mathcal{B} | \mathcal{A} \times \mathcal{B})} 2^{n I_{P_{A_i B_j} V_{A'B'|AB}}(A', B'; A, B)} \\ & \times \left[ \frac{\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{ij}^{pq} \right|}{4N_i M_{ij}} + \sum_{(k,l) \neq (i,j)} \frac{\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{kl} \right|}{4N_k M_{kl}} \right]. \end{aligned}$$

Immediately, normalizing by  $4N_i M_{ij}$  and taking the sum over  $1 \leq i \leq m_n$ ,  $1 \leq j \leq m'_{in}$ ,  $1 \leq p \leq N_i$ ,  $1 \leq q \leq M_{ij}$  yields

$$\mathbb{E} \sum_{i=1}^{m_n} \sum_{j=1}^{m'_{in}} \frac{1}{4N_i M_{ij}} \sum_{p=1}^{2N_i} \sum_{q=1}^{2M_{ij}} S_{ij}^{pq} \leq (n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n^2 (\max_i m'_{in})^2. \quad (113)$$

Similarly, it follows from (107) and (108) that

$$\mathbb{E} \sum_{i=1}^{m_n} \sum_{j=1}^{m'_{in}} \frac{1}{4N_i M_{ij}} \sum_{p=1}^{2N_i} \sum_{q=1}^{2M_{ij}} K_{ij}^{pq} \leq (n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n (\max_i m'_{in})^2 \leq (n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n^2 (\max_i m'_{in})^2, \quad (114)$$

where

$$\begin{aligned} K_{ij}^{pq} & \triangleq \sum_{V_{A'B'|AB} \in \mathcal{P}_n(\mathcal{A} \times \mathcal{B} | \mathcal{A} \times \mathcal{B})} 2^{n I_{P_{A_i B_j} V_{A'B'|AB}}(B'; B | A)} \\ & \times \left[ \frac{\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{ij}^{pq} \right|}{4M_{ij}} + \sum_{l \neq j} \frac{\left| \mathbb{T}_{V_{A'B'|AB}} \left( \mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)} \right) \cap \mathcal{C}_{il} \right|}{4M_{il}} \right], \end{aligned}$$

and it follows from (109) and (111) that

$$\mathbb{E} \sum_{i=1}^{m_n} \sum_{j=1}^{m'_{in}} \frac{1}{4N_i M_{ij}} \sum_{p=1}^{2N_i} \sum_{q=1}^{2M_{ij}} L_{ij}^{pq} \leq (n+1)^{2|\mathcal{A}|^2} m_n^2 (\max_i m'_{in}) \leq (n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n (\max_i m'_{in})^2, \quad (115)$$

where  $L_{ij}^{pq}$  is actually independent of  $j$  and  $q$  and is given by

$$L_{ij}^{pq} = L_i^p \triangleq \sum_{V_{A'|A} \in \mathcal{P}_n(\mathcal{A}|\mathcal{A})} 2^{n I_{P_{A_i} V_{A'|A}}(A'; A)} \\ \times \left[ \frac{|\mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \mathcal{C}_i^p|}{2N_i} + \sum_{k \neq i} \frac{|\mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \mathcal{C}_k|}{2N_k} \right].$$

Summing (113), (114) and (115) together, we obtain

$$\mathbb{E} \sum_{i=1}^{m_n} \sum_{j=1}^{m'_{in}} \frac{1}{4N_i M_{ij}} \sum_{p=1}^{2N_i} \sum_{q=1}^{2M_{ij}} (S_{ij}^{pq} + K_{ij}^{pq} + L_{ij}^{pq}) \leq 3(n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n^2 (\max_i m'_{in})^2. \quad (116)$$

Therefore, there exists at least a selection of these sets  $\{\widehat{\mathcal{C}}_i\}_{i=1}^{m_n}$  and  $\{\widehat{\mathcal{C}}_{ij}\}_{i=1, j=1}^{m_n, m'_{in}}$  such that

$$\sum_{i=1}^{m_n} \sum_{j=1}^{m'_{in}} \frac{1}{4N_i M_{ij}} \sum_{p=1}^{2N_i} \sum_{q=1}^{2M_{ij}} (S_{ij}^{pq} + K_{ij}^{pq} + L_{ij}^{pq}) \leq 3(n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n^2 (\max_i m'_{in})^2,$$

which implies that for all  $i = 1, 2, \dots, m_n$  and  $j = 1, 2, \dots, m'_{in}$  the following is satisfied

$$\frac{1}{4N_i M_{ij}} \sum_{p=1}^{2N_i} \sum_{q=1}^{2M_{ij}} (S_{ij}^{pq} + K_{ij}^{pq} + L_{ij}^{pq}) \leq 3(n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n^2 (\max_i m'_{in})^2. \quad (117)$$

We next proceed with an expurgation argument. Without loss of generality, we assume

$$\frac{1}{2M_{ij}} \sum_{q=1}^{2M_{ij}} (S_{ij}^{1q} + K_{ij}^{1q} + L_{ij}^{1q}) \leq \frac{1}{2M_{ij}} \sum_{q=1}^{2M_{ij}} (S_{ij}^{2q} + K_{ij}^{2q} + L_{ij}^{2q}) \leq \dots \\ \leq \frac{1}{2M_{ij}} \sum_{q=1}^{2M_{ij}} (S_{ij}^{2N_i, q} + K_{ij}^{2N_i, q} + L_{ij}^{2N_i, q}),$$

then we must have, for every  $1 \leq p \leq N_i$ ,

$$\frac{1}{2M_{ij}} \sum_{q=1}^{2M_{ij}} S_{ij}^{pq} + K_{ij}^{pq} + L_{ij}^{pq} \leq 6(n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n^2 (\max_i m'_{in})^2.$$

Similarly, suppose for each  $p = 1, 2, \dots, N_i$ ,

$$S_{ij}^{p1} + K_{ij}^{p1} + L_{ij}^{p1} \leq S_{ij}^{p2} + K_{ij}^{p2} + L_{ij}^{p2} \leq \dots \leq S_{ij}^{p, 2M_{ij}} + K_{ij}^{p, 2M_{ij}} + L_{ij}^{p, 2M_{ij}},$$

the above implies that for each  $p = 1, 2, \dots, N_i$  and each  $q = 1, 2, \dots, M_{ij}$ ,

$$S_{ij}^{pq} + K_{ij}^{pq} + L_{ij}^{pq} \leq 12(n+1)^{2|\mathcal{A}|^2 |\mathcal{B}|^2} m_n^2 (\max_i m'_{in})^2. \quad (118)$$

We now let for  $i = 1, 2, \dots, m_n$ ,  $p = 1, 2, \dots, N_i$ ,  $\Omega_i \triangleq \{\mathbf{a}_1^{(i)}, \mathbf{a}_2^{(i)}, \dots, \mathbf{a}_{N_i}^{(i)}\} \subseteq \widehat{\mathcal{C}}_i$ ,  $\Omega_i^p \triangleq \Omega_i / \{\mathbf{a}_p^{(i)}\} \subseteq \widehat{\mathcal{C}}_i^p$  and for  $j = 1, 2, \dots, m'_{in}$ ,  $q = 1, 2, \dots, M_{ij}$ , let  $\Omega_{ij}(\mathbf{a}_p^{(i)}) = \left\{ (\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \right\}_{q=1}^{M_{ij}}$  such that

$$\begin{aligned} \Omega_{ij} \triangleq \bigcup_{p=1}^{N_i} \Omega_{ij}(\mathbf{a}_p^{(i)}) &= \left\{ (\mathbf{a}_1^{(i)}, \mathbf{b}_{11}^{(j)}), (\mathbf{a}_1^{(i)}, \mathbf{b}_{12}^{(j)}), \dots, (\mathbf{a}_1^{(i)}, \mathbf{b}_{1,M_{ij}}^{(j)}), \right. \\ &\quad \left. (\mathbf{a}_2^{(i)}, \mathbf{b}_{21}^{(j)}), (\mathbf{a}_2^{(i)}, \mathbf{b}_{22}^{(j)}), \dots, (\mathbf{a}_2^{(i)}, \mathbf{b}_{2,M_{ij}}^{(j)}), \right. \\ &\quad \dots \dots \\ &\quad \left. (\mathbf{a}_{N_i}^{(i)}, \mathbf{b}_{N_i,1}^{(j)}), (\mathbf{a}_{N_i}^{(i)}, \mathbf{b}_{N_i,2}^{(j)}), \dots, (\mathbf{a}_{N_i}^{(i)}, \mathbf{b}_{N_i,M_{ij}}^{(j)}) \right\} \subseteq \widehat{\mathcal{C}}_{ij}, \end{aligned}$$

and denote also  $\Omega_{ij}^{pq} \triangleq \Omega_{ij} / \{(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)})\} \subseteq \widehat{\mathcal{C}}_{ij}^{pq}$ . Immediately, it follows from (118) that for every  $i = 1, 2, \dots, m_n$ ,  $j = 1, 2, \dots, m'_{in}$ ,  $k = 1, 2, \dots, m_n$ ,  $l = 1, 2, \dots, m'_{kn}$ ,  $p = 1, 2, \dots, N_i$ ,  $q = 1, 2, \dots, M_{ij}$ , and every  $V_{A'B'|AB} \in \mathcal{P}_n(\mathcal{A} \times \mathcal{B} | \mathcal{A} \times \mathcal{B})$  and  $V_{A'|A} \in \mathcal{P}_n(\mathcal{A} | \mathcal{A})$

$$\frac{\left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{kl} \right|}{N_k M_{kl}} \leq 2^{-n} \left[ I_{P_{A_i B_j} V_{A'B'|AB}}(A', B'; A, B) - \delta \right], \quad (k, l) \neq (i, j), \quad (119)$$

$$\frac{\left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{ij}^{pq} \right|}{N_i M_{ij}} \leq 2^{-n} \left[ I_{P_{A_i B_j} V_{A'B'|AB}}(A', B'; A, B) - \delta \right], \quad (120)$$

$$\frac{\left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{il} \right|}{M_{il}} \leq 2^{-n} \left[ I_{P_{A_i B_j} V_{A'B'|AB}}(B'; B | A) - \delta \right], \quad l \neq j, \quad (121)$$

$$\frac{\left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{ij}^{pq} \right|}{M_{ij}} \leq 2^{-n} \left[ I_{P_{A_i B_j} V_{A'B'|AB}}(B'; B | A) - \delta \right], \quad (122)$$

$$\frac{\left| \mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \Omega_k \right|}{N_k} \leq 2^{-n} \left[ I_{P_{A_i} V_{A'|A}}(A'; A) - \delta \right], \quad k \neq i, \quad (123)$$

$$\frac{\left| \mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \Omega_i^p \right|}{N_i} \leq 2^{-n} \left[ I_{P_{A_i} V_{A'|A}}(A'; A) - \delta \right], \quad (124)$$

where

$$\delta = \frac{2}{n} \left[ |\mathcal{A}|^2 |\mathcal{B}|^2 \log_2(n+1) + \log_2 m_n + \log_2(\max_i m'_{in}) + \log_2 12 \right].$$

Thus far, we proved the existence of the sets  $\Omega_i$  and  $\Omega_{ij}$  with elements selected uniformly from each  $\mathbb{T}_{A_i}$  and  $\mathbb{T}_{A_i B_j}$  satisfying the inequalities (119)–(124) for any  $V_{A'|A}$  and  $V_{A'B'|AB}$ . It remains to show that these sets are disjoint and have distinct elements provided assumptions (3) and (4). Indeed, since (123) and (124) hold for every  $V_{A'|A} \in \mathcal{P}_n(\mathcal{A} | \mathcal{A})$ , they of course hold when  $V_{A'|A}$  is a conditional distribution such that  $V_{A'|A}^*(a' | a)$  is 1 if  $a' = a$  and 0 otherwise. It then follows from (3)

$$\frac{1}{n} \log_2 N_i < H_{P_{A_i}}(A) - \delta = I_{P_{A_i} V_{A'|A}^*}^*(A'; A) - \delta$$

that  $\left| \mathbb{T}_{V_{A'|A}^*}(\mathbf{a}_p^{(i)}) \cap \Omega_k \right| = \left| \{\mathbf{a}_p^{(i)}\} \cap \Omega_k \right| < 1$  or equivalently,  $\left| \{\mathbf{a}_p^{(i)}\} \cap \Omega_k \right| = 0$ , which means any elements in  $\Omega_i$  does not belong to  $\Omega_k$  for  $i \neq k$ , i.e.,  $\Omega_i$  and  $\Omega_k$  are disjoint. Likewise, using assumption (3) in (124),

we see that

$$\left| \mathbb{T}_{V_{A'|A}^*}(\mathbf{a}_p^{(i)}) \cap \Omega_i^p \right| = \left| \{\mathbf{a}_p^{(i)}\} \cap \Omega_i^p \right| = 0,$$

which means that  $\Omega_i$  has  $N_i$  disjoint elements. Similarly, setting  $V_{A'B'|AB}$  be the conditional distribution such that  $V_{A'B'|AB}^*(a', b'|a, b)$  is 1 if  $a' = a$ ,  $b' = b$  and 0 otherwise, and using (4)

$$\frac{1}{n} \log_2 M_{ij} < H_{P_{A_i} P_{B_j|A_i}}(B|A) - \delta,$$

we see that for any  $\mathbf{a}_p^{(i)} \in \Omega_i$ ,  $\Omega_{ij}(\mathbf{a}_p^{(i)})$ 's are disjoint and the elements in  $\Omega_{ij}(\mathbf{a}_p^{(i)})$  are all distinct, i.e.,  $|\Omega_{ij}(\mathbf{a}_p^{(i)})| = M_{ij}$  for every  $\mathbf{a}^{(i)} \in \Omega_i$ . Finally, when  $V_{A'|A}$  is not the conditional distribution such that  $V_{A'|A}(a'|a)$  is 1 if  $a' = a$  and 0 otherwise, we can write (123) and (124) in the same way as (5), and when  $V_{A'B'|AB}$  is not the conditional distribution such that  $V_{A'B'|AB}(a', b'|a, b)$  is 1 if  $a' = a$ ,  $b' = b$  and 0 otherwise, we can write (119)–(120) as (6), and write (121)–(122) as (7), since

$$\begin{aligned} \left| \mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \Omega_i^p \right| &= \left| \mathbb{T}_{V_{A'|A}}(\mathbf{a}_p^{(i)}) \cap \Omega_i \right|, \\ \left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{ij}^{pq} \right| &= \left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{ij} \right|, \\ \left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{ij}^{pq} \right| &= \left| \mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}_p^{(i)}, \mathbf{b}_{p,q}^{(j)}) \cap \Omega_{ij} \right|. \end{aligned}$$

■

## B Proof of (26) and (27)

### B.1 Upper Bound on $\left| \mathbb{T}_{\widehat{V}_Y|TUX}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_1 \right|$

If we fix a  $k = 1, 2, \dots, m_n$  and a  $l = 1, 2, \dots, m'_{kn}$ , then  $\mathcal{E}_1$  is the set of all  $\mathbf{y}$  such that there exist some  $((\mathbf{t}, \mathbf{u})', \mathbf{x}') \in \Omega_{kl}$ ,  $(\mathbf{t}, \mathbf{u})' \neq (\mathbf{t}, \mathbf{u})$ ,  $((\mathbf{t}, \mathbf{u}), \mathbf{x}, (\mathbf{t}, \mathbf{u})', \mathbf{x}', \mathbf{y})$  admits a joint type  $P_{(\mathbf{t}, \mathbf{u})\mathbf{x}(\mathbf{t}, \mathbf{u})'\mathbf{x}'\mathbf{y}} \in \mathcal{P}_n(\mathcal{T}^2 \times \mathcal{U}^2 \times \mathcal{X}^2 \times \mathcal{Y})$  and

$$I((\mathbf{t}, \mathbf{u})', \mathbf{x}'; \mathbf{y}) - (R_k + R_{kl}) \geq I((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{y}) - (R_i + R_{ij}). \quad (125)$$

Note that (125) can be represented as for dummy R.V.'s  $(TU) \in \mathcal{T} \times \mathcal{U}$ ,  $X \in \mathcal{X}$ ,  $(TU)' \in \mathcal{T} \times \mathcal{U}$ ,  $X' \in \mathcal{X}$ , and  $Y \in \mathcal{Y}$ , the following holds under the joint distribution  $P_{(TU)X(TU)'X'Y} = P_{(\mathbf{t}, \mathbf{u})\mathbf{x}(\mathbf{t}, \mathbf{u})'\mathbf{x}'\mathbf{y}}$ ,

$$I_{P_{(TU)'X'Y}}((T, U)', X'; Y) - (R_k + R_{kl}) \geq I_{P_{TUXY}}((T, U), X; Y) - (R_i + R_{ij}),$$

where  $P_{(TU)'X'Y}$  and  $P_{TUXY}$  are the corresponding marginal distributions induced by  $P_{(TU)X(TU)'X'Y}$ . Thus,  $\mathbb{T}_{\widehat{V}_Y|TUX}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_1$  can be written as a union of subsets

$$\mathbb{T}_{\widehat{V}_Y|(TU)X}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_1 = \bigcup_{k=1}^{m_n} \bigcup_{l=1}^{m'_{kn}} \bigcup_{P_{(TU)X(TU)'X'Y} \in \mathcal{C}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x})} \mathcal{F}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)X(TU)'X'Y}) \quad (126)$$

where

$$\mathcal{C}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \triangleq \left\{ \begin{array}{l} P_{(TU)X} = P_{(\mathbf{t}, \mathbf{u})\mathbf{x}} = P_{(TU)_i X_j}, \\ P_{(TU)X(TU)'X'Y} \\ \in \mathcal{P}_n(\mathcal{T}^2 \times \mathcal{U}^2 \times \mathcal{X}^2 \times \mathcal{Y}) : \quad \begin{array}{l} P_{(TU)'X'} = P_{(TU)_k X_l}, \quad P_{Y|(TU)X} = \widehat{V}_{Y|(TU)X}, \\ I_{P_{(TU)'X'Y}}((T, U)', X'; Y) - (R_k + R_{kl}) \\ \geq I_{P_{(TU)XY}}((T, U), X; Y) - (R_i + R_{ij}) \end{array} \end{array} \right\},$$

where  $P_{(TU)X}$ ,  $P_{(TU)'X'}$  and  $P_{Y|(TU)X}$ , etc, are the corresponding marginal and conditional distributions induced by  $P_{(TU)X(TU)'X'Y}$ , and

$$\mathcal{F}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)X(TU)'X'Y}) \triangleq \left\{ \mathbf{y} : \begin{array}{l} \exists ((\mathbf{t}, \mathbf{u})', \mathbf{x}') \quad ((\mathbf{t}, \mathbf{u}), \mathbf{x}, (\mathbf{t}, \mathbf{u})', \mathbf{x}', \mathbf{y}) \in \mathbb{T}_{(TU)X(TU)'X'Y} \\ \text{such that} \quad ((\mathbf{t}, \mathbf{u})', \mathbf{x}') \in \Omega_{kl}, \quad (\mathbf{t}, \mathbf{u})' \neq (\mathbf{t}, \mathbf{u}) \end{array} \right\},$$

where  $\mathbb{T}_{(TU)X(TU)'X'Y} \triangleq \mathbb{T}_{P_{(TU)X(TU)'X'Y}}$ . Clearly, given any  $k, l$ , and  $P_{(TU)X(TU)'X'Y}$ ,

$$\begin{aligned} & |\mathcal{F}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)X(TU)'X'Y})| \\ & \leq \left| \left\{ ((\mathbf{t}, \mathbf{u})', \mathbf{x}', \mathbf{y}) : \begin{array}{l} ((\mathbf{t}, \mathbf{u}), \mathbf{x}, (\mathbf{t}, \mathbf{u})', \mathbf{x}', \mathbf{y}) \in \mathbb{T}_{(TU)X(TU)'X'Y} \\ ((\mathbf{t}, \mathbf{u})', \mathbf{x}') \in \Omega_{kl}, \quad (\mathbf{t}, \mathbf{u})' \neq (\mathbf{t}, \mathbf{u}) \end{array} \right\} \right| \\ & = \left| \left\{ ((\mathbf{t}, \mathbf{u})', \mathbf{x}') : \begin{array}{l} ((\mathbf{t}, \mathbf{u}), \mathbf{x}, (\mathbf{t}, \mathbf{u})', \mathbf{x}') \in \mathbb{T}_{(TU)X(TU)'X'Y} \\ ((\mathbf{t}, \mathbf{u})', \mathbf{x}') \in \Omega_{kl}, \quad (\mathbf{t}, \mathbf{u})' \neq (\mathbf{t}, \mathbf{u}) \end{array} \right\} \right| \times |\mathbb{T}_{Y|(TU)X(TU)'X'}((\mathbf{t}, \mathbf{u}), \mathbf{x}, (\mathbf{t}, \mathbf{u})', \mathbf{x}')| \\ & \leq N_k M_{kl} 2^{-n} \left[ I_{P_{(TU)X(TU)'X'Y}}((T, U), X; (T, U)', X') - \eta \right] \times 2^{nH_{P_{(TU)X(TU)'X'Y}}(Y|(T, U), X, (T, U)', X')}, \end{aligned} \quad (127)$$

where the last inequality follows from Lemma 3. Meanwhile, when  $((\mathbf{t}, \mathbf{u}), \mathbf{x}) \in \Omega_{ij}$ , the following simple bound also holds

$$|\mathcal{F}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)X(TU)'X'Y})| \leq |\mathbb{T}_{Y|(TU)X}((\mathbf{t}, \mathbf{u}), \mathbf{x})| \leq 2^{nH_{P_{(TU)XY}}(Y|(T, U), X)} = 2^{nH_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}}(Y|(T, U), X))}} \quad (128)$$

since for each  $\mathbb{T}_{(TU)X(TU)'X'Y} \in \mathcal{C}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x})$ , we have  $P_{(TU)X} = P_{((TU)_i X_j}$ ,  $P_{Y|(TU)X} = \widehat{V}_{Y|(TU)X}$  and hence  $P_{(TU)XY} = P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}$ . Now substituting the following inequality (cf. [8, Eq. (28)])

$$\begin{aligned} & H_{P_{(TU)X(TU)'X'Y}}(Y|(T, U), X, (T, U)', X') - I_{P_{(TU)X(TU)'X'Y}}((T, U), X; (T, U)', X') \\ & = H_{P_{(TU)XY}}(Y|(T, U), X) - I_{P_{(TU)X(TU)'X'Y}}((T, U)', X'; (T, U), X, Y) \\ & \leq H_{P_{(TU)XY}}(Y|(T, U), X) - I_{P_{(TU)'X'Y}}((T, U)', X'; Y) \end{aligned} \quad (129)$$

into (127), combining with (128) together, we obtain

$$|\mathcal{F}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)X(TU)'X'Y})| \leq 2^n \left[ H_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}}(Y|(T, U), X) - \left| I_{P_{(TU)'X'Y}}((T, U)', X'; Y) - (R_k + R_{kl}) \right|^+ \right]. \quad (130)$$

Again recall that for  $P_{(TU)X(TU)'X'Y} \in \mathcal{C}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x})$ ,  $P_{(TU)XY} = P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}$ , and note that

$$I_{P_{(TU)'X'Y}}((T, U)', X'; Y) - (R_k + R_{kl}) \geq I_{P_{(TU)XY}}((T, U), X; Y) - (R_i + R_{ij}).$$

This implies when  $P_{(TU)X(TU)'X'Y} \in \mathcal{C}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x})$

$$|\mathcal{F}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)X(TU)'X'Y})| \leq 2^n \left[ H_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}^{(Y|(T,U),X)}} - \left| I_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}^{((T,U),X;Y)-(R_i+R_{ij})}} \right|^+ \right],$$

and hence

$$\begin{aligned} \left| \mathbb{T}_{\widehat{V}_{Y|(TU)X}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_1 \right| &\leq m_n \left( \max_i m'_{in} \right) (n+1)^{|\mathcal{T} \times \mathcal{U}|^2 |\mathcal{X}|^2 |\mathcal{Y}|} \\ &\times 2^n \left[ H_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}^{(Y|(T,U),X)}} - \left| I_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}^{((T,U),X;Y)-(R_i+R_{ij})}} \right|^+ \right], \end{aligned}$$

since by Lemma 1

$$|\mathcal{C}_{k,l}((\mathbf{t}, \mathbf{u}), \mathbf{x})| \leq |\mathcal{P}_n(\mathcal{T}^2 \times \mathcal{U}^2 \times \mathcal{X}^2 \times \mathcal{Y})| \leq (n+1)^{|\mathcal{T}|^2 |\mathcal{U}|^2 |\mathcal{X}|^2 |\mathcal{Y}|}.$$

## B.2 Upper Bound on $\left| \mathbb{T}_{\widehat{V}_{Y|(TU)X}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_2 \right|$

If we fix an  $i = 1, 2, \dots, m_n$  and an  $l = 1, 2, \dots, m'_{in}$ , then  $\mathcal{E}_2$  is the set of all  $\mathbf{y}$  such that there exist some  $((\mathbf{t}, \mathbf{u}), \mathbf{x}') \in \Omega_{il}$ ,  $\mathbf{x}' \neq \mathbf{x}$ ,  $((\mathbf{t}, \mathbf{u}), \mathbf{x}, \mathbf{x}', \mathbf{y})$  admits a joint type  $P_{(\mathbf{t}, \mathbf{u})\mathbf{x}\mathbf{x}'\mathbf{y}} \in \mathcal{P}_n(\mathcal{T} \times \mathcal{U} \times \mathcal{X}^2 \times \mathcal{Y})$  and

$$I((\mathbf{t}, \mathbf{u}), \mathbf{x}'; \mathbf{y}) - (R_i + R_{il}) \geq I((\mathbf{t}, \mathbf{u}), \mathbf{x}; \mathbf{y}) - (R_i + R_{ij}). \quad (131)$$

Using the identity

$$I((T, U), X; Y) = I(T, U; Y) + I(X; Y|T, U),$$

on both sides of (131) we see it is equivalent to

$$I(\mathbf{x}'; \mathbf{y}|\mathbf{t}, \mathbf{u}) - R_{il} \geq I(\mathbf{x}; \mathbf{y}|\mathbf{t}, \mathbf{u}) - R_{ij}. \quad (132)$$

Note that (132) can be represented as for dummy R.V.'s  $(TU) \in \mathcal{T} \times \mathcal{U}$ ,  $X \in \mathcal{X}$ ,  $X' \in \mathcal{X}$ , and  $Y \in \mathcal{Y}$ , the following holds under the joint distribution  $P_{(TU)XX'Y} = P_{(\mathbf{t}, \mathbf{u})\mathbf{x}\mathbf{x}'\mathbf{y}}$ ,

$$I_{P_{(TU)X'Y}}(X'; Y|T, U) - R_{il} \geq I_{P_{(TU)XY}}(X; Y|T, U) - R_{ij},$$

where  $P_{(TU)XY}$  and  $P_{(TU)X'Y}$  are the corresponding marginal distributions induced by  $P_{(TU)XX'Y}$ . Thus,  $\mathbb{T}_{\widehat{V}_{Y|(TU)X}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_2$  can be written as a union of subsets

$$\mathbb{T}_{\widehat{V}_{Y|(TU)X}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_2 = \bigcup_{l=1}^{m'_{in}} \bigcup_{P_{(TU)XX'Y} \in \mathcal{C}_l((\mathbf{t}, \mathbf{u}), \mathbf{x})} \mathcal{F}_l((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)XX'Y}) \quad (133)$$

where

$$\mathcal{C}_l((\mathbf{t}, \mathbf{u}), \mathbf{x}) \triangleq \left\{ \begin{array}{l} P_{(TU)X} = P_{(\mathbf{t}, \mathbf{u})\mathbf{x}} = P_{(TU)_i X_j}, \\ P_{(TU)X'Y} \\ \in \mathcal{P}_n(\mathcal{T} \times \mathcal{U} \times \mathcal{X}^2 \times \mathcal{Y}) : \quad \begin{array}{l} P_{(TU)X'} = P_{(TU)_i X_l}, \quad P_{Y|(TU)X} = \widehat{V}_{Y|TUX} \\ I_{P_{(TU)X'Y}}(X'; Y|T, U) - R_{il} \\ \geq I_{P_{(TU)XY}}(X; Y|T, U) - R_{ij} \end{array} \end{array} \right\},$$

where  $P_{(TU)X}$ ,  $P_{(TU)X'}$  and  $P_{Y|(TU)X}$ , etc, are the corresponding marginal and conditional distributions induced by  $P_{(TU)XX'Y}$ , and

$$\mathcal{F}_l((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)XX'Y}) \triangleq \left\{ \mathbf{y} : \begin{array}{l} \exists ((\mathbf{t}, \mathbf{u}), \mathbf{x}') \quad ((\mathbf{t}, \mathbf{u}), \mathbf{x}, \mathbf{x}', \mathbf{y}) \in \mathbb{T}_{(TU)XX'Y} \\ \text{such that} \quad ((\mathbf{t}, \mathbf{u}), \mathbf{x}') \in \Omega_{il}, \quad \mathbf{x}' \neq \mathbf{x} \end{array} \right\},$$

where  $\mathbb{T}_{(TU)XX'Y} = \mathbb{T}_{P_{(TU)XX'Y}}$ . Using a similar counting argument, and applying Lemma 3, we can bound, for any  $l = 1, 2, \dots, m'_{in}$  and  $P_{(TU)XX'Y} \in \mathcal{C}_l((\mathbf{t}, \mathbf{u}), \mathbf{x})$ ,

$$|\mathcal{F}_l((\mathbf{t}, \mathbf{u}), \mathbf{x}, P_{(TU)XX'Y})| \leq 2^n \left[ H_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}}(Y|(T,U), X)} - \left| I_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}}(X; Y|T, U) - R_{ij} \right|^+ \right],$$

and finally, we obtain,

$$\begin{aligned} \left| \mathbb{T}_{\widehat{V}_{Y|(TU)X}}((\mathbf{t}, \mathbf{u}), \mathbf{x}) \cap \mathcal{E}_2 \right| &\leq \left( \max_i m'_{in} \right) (n+1)^{|T \times \mathcal{U}| |\mathcal{X}|^2 |\mathcal{Y}|} \\ &\times 2^n \left[ H_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}}(Y|(T,U), X)} - \left| I_{P_{((TU)_i X_j} \widehat{V}_{Y|(TU)X}}(X; Y|T, U) - R_{ij} \right|^+ \right] \end{aligned}$$

since  $|\mathcal{C}_l((\mathbf{t}, \mathbf{u}), \mathbf{x})| \leq (n+1)^{|T| |\mathcal{U}| |\mathcal{X}|^2 |\mathcal{Y}|}$ . ■

## C Proof of Theorem 3

*Forward Part (1):* It follows from (19)-(21) that  $E_r(R_1, R_2, W_{YZ|TUX}, P_{TUX}) > 0$  if and only if  $(R_1, R_2) \in \mathcal{R}(W_{YZ|TUX}, P_{TUX})$ . It then follows that  $E_r(R_1, R_2, W_{YZ|UX}) > 0$  if  $(R_1, R_2) \in \mathcal{R}(W_{YZ|UX})$ . According to Theorem 2 and the definition of the system JSCC error exponent,  $P_e^{(n)} \rightarrow 0$  if the lower bound (34) is positive, which needs  $E_r(\tau H_P(S), \tau H_P(L|S), W_{YZ|UX}) > 0$ . This means  $P_e^{(n)} \rightarrow 0$  if the pair  $(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}(W_{YZ|UX})$ .

*Converse Part (2):* The proof follows from a similar manner as the converse part of [16, Theorem 1] for a broadcast channel. For the sake of completeness, we also provide a full proof here since we deal with a 2-user channel. We first prove the following lemma.

### Lemma 7

$$\overline{\mathcal{R}}(W_{YZ|UX}) = \mathcal{R}'(W_{YZ|UX}),$$

where  $\mathcal{R}'(W_{YZ|UX})$  is defined in (49).



**Proof:** It is straightforward to see that  $\overline{\mathcal{R}}(W_{YZ|UX}) \subseteq \mathcal{R}'(W_{YZ|UX})$ . To complete the proof, it suffices to show  $\mathcal{R}'(W_{YZ|UX}) \subseteq \overline{\mathcal{R}}(W_{YZ|UX})$ . We note that both  $\overline{\mathcal{R}}(W_{YZ|UX})$  and  $\mathcal{R}'(W_{YZ|UX})$  are convex and closed. Therefore, instead of verifying that all  $(R_1, R_2)$ 's in  $\mathcal{R}'(W_{YZ|UX})$  belong to  $\overline{\mathcal{R}}(W_{YZ|UX})$ , we show that all the boundary points of  $\mathcal{R}'(W_{YZ|UX})$  are in  $\overline{\mathcal{R}}(W_{YZ|UX})$ . By the definition of  $\mathcal{R}'(W_{YZ|UX})$ , we note that any boundary point  $(R_1, R_2)$  of  $\mathcal{R}'(W_{YZ|UX})$  has to satisfy at least one of the conditions:

- Case 1: there exist RV  $T$  and  $P_{TUX}$  such that

$$\begin{aligned} I(U, X; Y) &\leq I(X; Y|T, U) + I(T, U; Z) \\ R_1 + R_2 &= I(U, X; Y) \\ R_1 &= I(T, U; Z). \end{aligned}$$

This is true since if  $R_1 + R_2 < I(U, X; Y)$  or  $R_1 < I(T, U; Z)$ , we can increase  $R_1$  or  $R_2$  which contradicts the boundary point assumption on  $(R_1, R_2)$ .

- Case 2: there exist RV  $T$  and  $P_{TUX}$  such that

$$\begin{aligned} I(U, X; Y) &\geq I(X; Y|T, U) + I(T, U; Z) \\ R_1 + R_2 &= I(X; Y|T, U) + I(T, U; Z) \\ R_1 &= I(T, U; Z). \end{aligned}$$

Now if the boundary point  $(R_1, R_2)$  satisfies Case 1, clearly, for the same  $T$  and  $P_{TUX}$ , we have

$$\begin{aligned} R_1 + R_2 &= I(U, X; Y) \\ R_1 &= I(T, U; Z) \\ R_2 &\leq I(X; Y|T, U). \end{aligned}$$

This shows that  $(R_1, R_2) \in \overline{\mathcal{R}}(W_{YZ|UX})$ . Similarly, if the boundary point  $(R_1, R_2)$  satisfies Case 2, for such  $T$  and  $P_{TUX}$ , we have

$$\begin{aligned} R_2 &= I(X; Y|T, U) \\ R_1 &= I(T, U; Z) \\ R_1 + R_2 &\leq I(U, X; Y), \end{aligned}$$

and thus,  $(R_1, R_2) \in \overline{\mathcal{R}}(W_{YZ|UX})$ . Since the boundary points of  $\mathcal{R}'(W_{YZ|UX})$  are in  $\overline{\mathcal{R}}(W_{YZ|UX})$ , we conclude that the entire region of  $\mathcal{R}'(W_{YZ|UX})$  is in  $\overline{\mathcal{R}}(W_{YZ|UX})$ , and hence Lemma 7 is proved. ■

By lemma 7, it suffices to show that, for any  $\epsilon > 0$ , if

$$\max \left\{ P_{Y_e}^{(n)}(Q_{SL}, W_{YZ|XU}, \tau), P_{Z_e}^{(n)}(Q_{SL}, W_{YZ|UX}, \tau) \right\} \leq \epsilon_n \rightarrow 0$$

as  $n$  goes to infinity, then there exists a RV  $T$  satisfying  $T \rightarrow (U, X) \rightarrow (Y, Z)$ , i.e., the joint distribution  $P_{TUXYZ}$  can be factorized as  $P_T P_{UX|T} W_{YZ|UX}$ , such that  $(\tau H_Q(S), \tau H_Q(L|S)) \in \mathcal{R}'(W_{YZ|UX}, P_{TUX})$ , i.e.,

$$\begin{aligned}\tau H_Q(S, L) &\leq \min\{I(U, X; Y), I(X; Y|T, U) + I(T, U; Z)\}, \\ \tau H_Q(S) &\leq I(T, U; Z).\end{aligned}$$

Fix  $k = \tau n$ . Fano's inequality gives

$$H(S^k, L^k | Y^n) \leq P_{Y_e}^{(n)} \log_2 |\mathcal{S}^k \times \mathcal{L}^k| + H(P_{Y_e}^{(n)}) \triangleq n\epsilon_{1n} \quad (134)$$

$$H(S^k | Z^n) \leq P_{Z_e}^{(n)} \log_2 |\mathcal{S}^k| + H(P_{Z_e}^{(n)}) \triangleq n\epsilon_{2n}, \quad (135)$$

where  $S^k \triangleq (S_1, S_2, \dots, S_k)$ ; similar definitions apply for the other tuples. It follows from (134)-(135) that

$$\begin{aligned}kH(S, L) &= H(L^k | S^k) + H(S^k) \\ &= I(L^k; Y^n | S^k) + H(L^k | S^k, Y^n) + I(S^k; Z^n) + H(S^k | Z^n) \\ &\leq \sum_{i=1}^n [I(L^k; Y_i | S^k, Y^{i-1}) + I(S^k; Z_i | \mathbf{Z}^{i+1})] + H(S^k, L^k | Y^n) + n\epsilon_{2n} \\ &\leq \sum_{i=1}^n [I(L^k, \mathbf{Z}^{i+1}; Y_i | S^k, Y^{i-1}) + I(S^k, Y^{i-1}; Z_i | \mathbf{Z}^{i+1}) - I(Y^{i-1}; Z_i | S^k, \mathbf{Z}^{i+1})] + n(\epsilon_{1n} + \epsilon_{2n}), \\ &\leq \sum_{i=1}^n \left[ I(L^k; Y_i | S^k, Y^{i-1}, \mathbf{Z}^{i+1}) + I(\mathbf{Z}^{i+1}; Y_i | S^k, Y^{i-1}) \right. \\ &\quad \left. + I(S^k, \mathbf{Z}^{i+1}, Y^{i-1}; Z_i) - I(Y^{i-1}; Z_i | S^k, \mathbf{Z}^{i+1}) \right] + n(\epsilon_{1n} + \epsilon_{2n}),\end{aligned}$$

where  $Y^{i-1} = (Y_1, Y_2, \dots, Y_{i-1})$  and  $\mathbf{Z}^{i+1} \triangleq (Z_{i+1}, Z_{i+2}, \dots, Z_n)$ . Substituting the identity [11, Lemma 7]

$$\sum_{i=1}^n I(\mathbf{Z}^{i+1}; Y_i | S^k, Y^{i-1}) = \sum_{i=1}^n I(Y^{i-1}; Z_i | S^k, \mathbf{Z}^{i+1})$$

into the above, and setting  $T_i = (S^k, Y^{i-1}, \mathbf{Z}^{i+1})$  for  $1 \leq i \leq n$  yields

$$\begin{aligned}kH(S, L) &\leq \sum_{i=1}^n \left[ I(L^k; Y_i | T_i) + I(T_i; Z_i) \right] + n(\epsilon_{1n} + \epsilon_{2n}) \\ &\stackrel{(a)}{=} \sum_{i=1}^n \left[ I(L^k; Y_i | T_i, U_i) + I(T_i, U_i; Z_i) \right] + n(\epsilon_{1n} + \epsilon_{2n}) \\ &\stackrel{(b)}{\leq} \sum_{i=1}^n \left[ I(X^n; Y_i | T_i, U_i) + I(T_i, U_i; Z_i) \right] + n(\epsilon_{1n} + \epsilon_{2n}) \\ &\stackrel{(c)}{=} \sum_{i=1}^n \left[ I(X_i; Y_i | T_i, U_i) + I(T_i, U_i; Z_i) \right] + n(\epsilon_{1n} + \epsilon_{2n}),\end{aligned} \quad (136)$$

where (a) holds since  $U_i$  is a deterministic function of  $S^k$  and hence of  $T_i$ , (b) follows from the data processing inequality, and (c) holds since  $Y_i$  is only determined by  $U_i$  and  $X_i$  due to the memoryless property of the channel. On the other hand,  $kH(S, L)$  can also be bounded by

$$\begin{aligned}
kH(S, L) &= H(S^k, L^k) \\
&= I(S^k, L^k; Y^n) + H(S^k, L^k | Y^n) \\
&\leq I(X^n, U^n; Y^n) + n\epsilon_{1n} \\
&= \sum_{i=1}^n I(U_i, X_i; Y_i) + n\epsilon_{1n}.
\end{aligned} \tag{137}$$

Likewise, it follows from (135) that

$$\begin{aligned}
kH(S) &= H(S^k) \\
&= I(S^k; Z^n) + H(S^k | Z^n) \\
&= \sum_{i=1}^n I(S^k; Z_i | \mathbf{Z}^{i+1}) + H(S^k | Z^n) \\
&\leq \sum_{i=1}^n I(S^k, \mathbf{Z}^{i+1}; Z_i) + n\epsilon_{2n} \\
&\leq \sum_{i=1}^n I(S^k, Y^{i-1}, \mathbf{Z}^{i+1}, U_i; Z_i) + n\epsilon_{2n} \\
&= \sum_{i=1}^n I(T_i, U_i; Z_i) + n\epsilon_{2n}.
\end{aligned} \tag{138}$$

Note also that  $T_i \longrightarrow (U_i, X_i) \longrightarrow (Y_i, Z_i)$  for all  $1 \leq i \leq n$ . According to (136), (137), and (138), and recalling that  $k = \tau n$ , it is easy to show (e.g., see [11]) that there exists an auxiliary RV  $T$  with  $P_{TUXYZ} = P_T P_{UX|T} W_{YZ|UX}$  such that

$$\begin{aligned}
\tau H(S, L) &\leq \min \{ I_{P_{UXYZ}}(U, X; Y), I_{P_{TUXYZ}}(X; Y | T, U) + I_{P_{TUXYZ}}(T, U; Z) \} \\
\tau H(S) &\leq I_{P_{TUXYZ}}(T, U; Z).
\end{aligned}$$

It remains to show that the alphabet of the RV  $T$  can be limited by  $|\mathcal{T}| \leq |\mathcal{U}||\mathcal{X}| + 1$ ; i.e., we will show by applying the support lemma below, which is based on the Carathéodory theorem (cf. [10, p. 311]) that there exists a RV  $\hat{T}$  with  $|\hat{\mathcal{T}}| \leq |\mathcal{U}||\mathcal{X}| + 1$  such that  $P_{\hat{T}UXYZ} = P_{\hat{T}} P_{UX|\hat{T}} W_{YZ|UX}$  and

$$\begin{aligned}
&(I_{P_{UXYZ}}(U, X; Y), I_{P_{TUXYZ}}(T, U; Z), I_{P_{TUXYZ}}(X; Y | T, U)) \\
&= (I_{P_{\hat{T}UXYZ}}(U, X; Y), I_{P_{\hat{T}UXYZ}}(\hat{T}, U; Z), I_{P_{\hat{T}UXYZ}}(X; Y | \hat{T}, U)).
\end{aligned} \tag{139}$$

**Lemma 8** ([10, Support lemma, p. 311]) Let  $f_j$ ,  $j = 1, 2, \dots, k$  be real-valued continuous functions on  $\mathcal{P}(\mathcal{X})$ . For any probability measure  $\mu$  on the Borel  $\sigma$ -algebra of  $\mathcal{P}(\mathcal{X})$ , there exist  $k$  elements  $P_1, P_2, \dots, P_k$

of  $\mathcal{P}(\mathcal{X})$  and  $k$  non-negative reals  $\alpha_1, \alpha_2, \dots, \alpha_k$  with  $\sum_{i=1}^k \alpha_i = 1$  such that for every  $j = 1, 2, \dots, k$

$$\int_{\mathcal{P}(\mathcal{X})} f_j(P) \mu(dP) = \sum_{i=1}^k \alpha_i f_j(P_i).$$

We first rewrite

$$I_{P_{TUXYZ}}(T, U; Z) = H(Z) - H(Z|T, U)$$

and

$$I_{P_{TUXYZ}}(X; Y|T, U) = H(Y|T, U) - H(Y|X, T, U) = H(Y|T, U) - H(Y|X, U)$$

where the last equality follows since  $T \rightarrow (U, X) \rightarrow Y$  forms a Markov chain. To apply the support lemma, we define the following real-valued continuous functions of distribution  $P_{UX|T}(\cdot, \cdot|t)$  on  $\mathcal{P}(\mathcal{U} \times \mathcal{X})$  for fixed  $t \in \mathcal{T}$ ,

$$f_i(P_{UX|T}(u, x|t)) \triangleq P_{UX|T}(u, x|t)$$

for all  $(u, x) \in \mathcal{U} \times \mathcal{X}$  except one pair  $(u, x)$ , so there are  $m - 1 = |\mathcal{U}||\mathcal{X}| - 1$  functions; i.e.,  $i$  ranges from 1 to  $m$ . Furthermore, we define real-valued continuous functions

$$f_m(P_{UX|T}(u, x|t)) \triangleq H(Z|T = t, U)$$

and

$$f_{m+1}(P_{UX|T}(u, x|t)) \triangleq H(Y|T = t, U).$$

According to the support lemma, there must exist a new RV  $\hat{T}$  (jointly distributed with  $(U, X)$ ) with alphabet size  $|\hat{T}| = m + 1 = |\mathcal{U}||\mathcal{X}| + 1$  such that the expectation of  $f_i$  with respect to  $P_T$ ,  $i = 1, 2, \dots, m + 1$ , can be expressed in terms of the convex combination of  $m + 1$  points, i.e.,

$$P_{UX}(u, x) = \sum_T P_T(t) f_i(P_{UX|T}(\cdot, \cdot|t)) = \sum_{\hat{T}} P_{\hat{T}}(\hat{t}) f_i(P_{UX|\hat{T}}(\cdot, \cdot|\hat{t})), \quad i = 1, 2, \dots, m - 1 \quad (140)$$

$$H(Z|T, U) = \sum_T P_T(t) f_m(P_{UX|T}(\cdot, \cdot|t)) = \sum_{\hat{T}} P_{\hat{T}}(\hat{t}) f_m(P_{UX|\hat{T}}(\cdot, \cdot|\hat{t})) = H(Z|\hat{T}, U) \quad (141)$$

and

$$H(Y|T, U) = \sum_T P_T(t) f_{m+1}(P_{UX|T}(\cdot, \cdot|t)) = \sum_{\hat{T}} P_{\hat{T}}(\hat{t}) f_{m+1}(P_{UX|\hat{T}}(\cdot, \cdot|\hat{t})) = H(Y|\hat{T}, U). \quad (142)$$

Clearly  $\hat{T} \rightarrow (U, X) \rightarrow (Y, Z)$  forms a Markov chain and (139) holds. The proof for the converse part is complete.  $\blacksquare$

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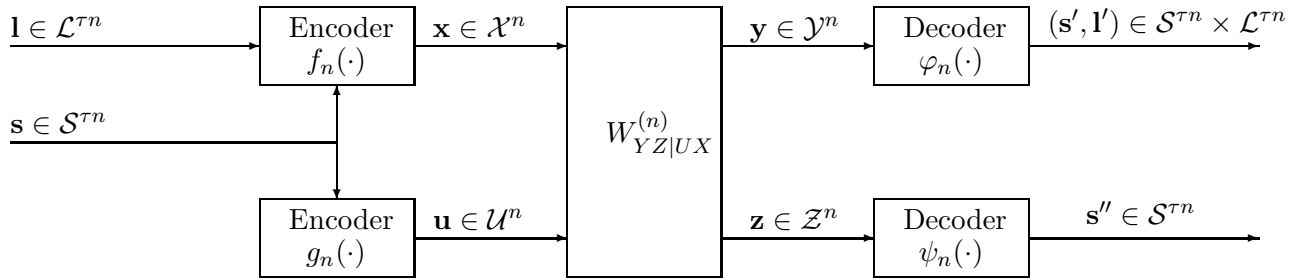


Figure 1: Transmitting two CS over the asymmetric 2-user communication channel.

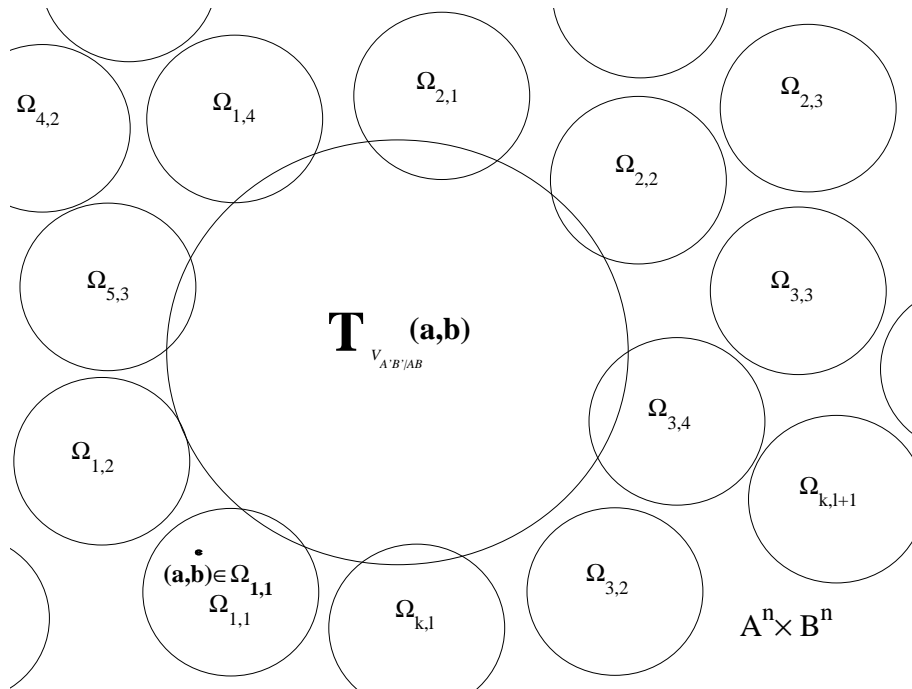


Figure 2: A graphical illustration of the (2-dimensional) joint type packing lemma (Lemma 3): there exist disjoint subsets  $\Omega_{ij}$ 's with bounded cardinalities in the “2-dimensional” space  $\mathcal{A}^n \times \mathcal{B}^n$  such that for any  $(\mathbf{a}, \mathbf{b}) \in \Omega_{ij}$  (say,  $(\mathbf{a}, \mathbf{b}) \in \Omega_{1,1}$ ), the size of the intersection between the sphere  $\mathbb{T}_{V_{A'B'|AB}}(\mathbf{a}, \mathbf{b})$  and every set  $\Omega_{kl}$  is “exponentially small” compared with the size of each  $\Omega_{kl}$ .



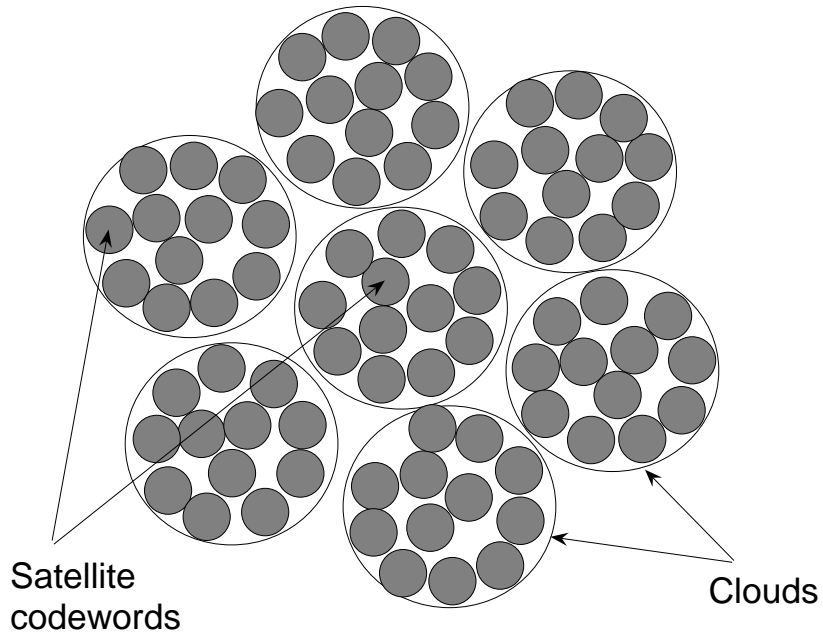


Figure 3: Relation between clouds and satellite codewords in superposition coding.

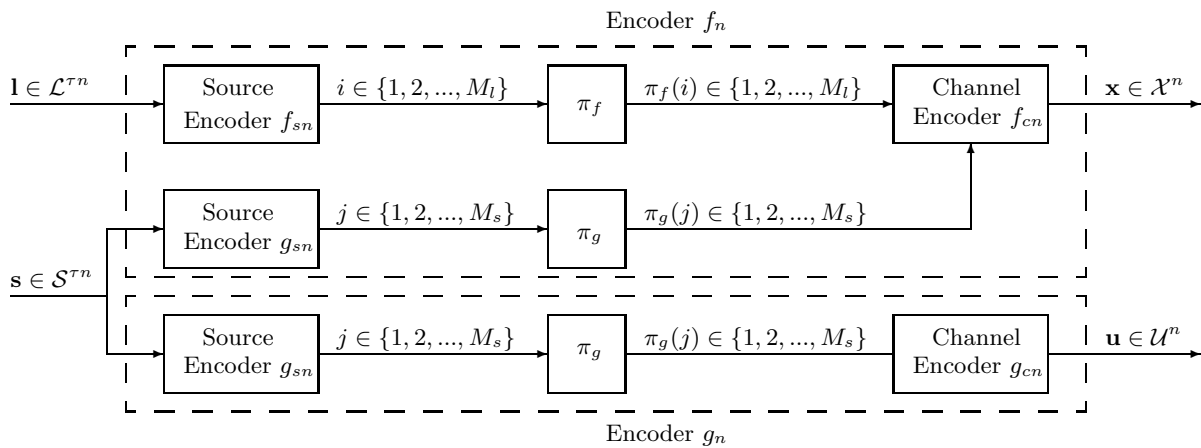


Figure 4: Tandem source-channel coding system - encoders.

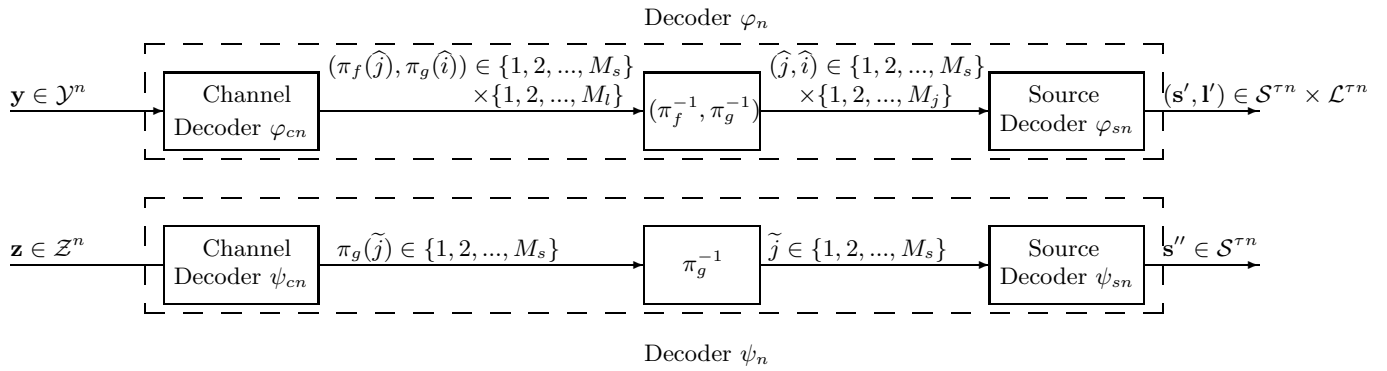


Figure 5: Tandem source-channel coding system - decoders.

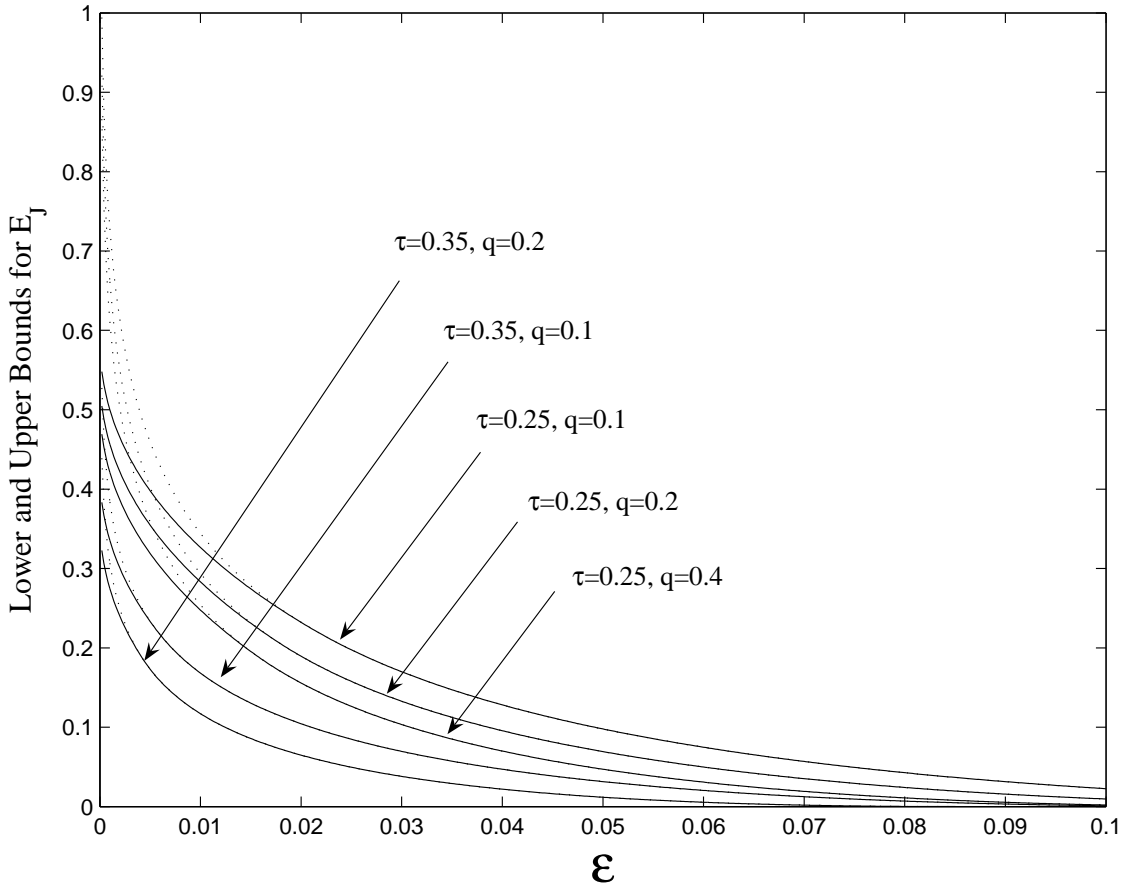


Figure 6: The lower bound (solid line) and the upper bound (dash line) for the system JSCC error exponent for transmitting binary CS over the binary AMAC with binary additive noise.

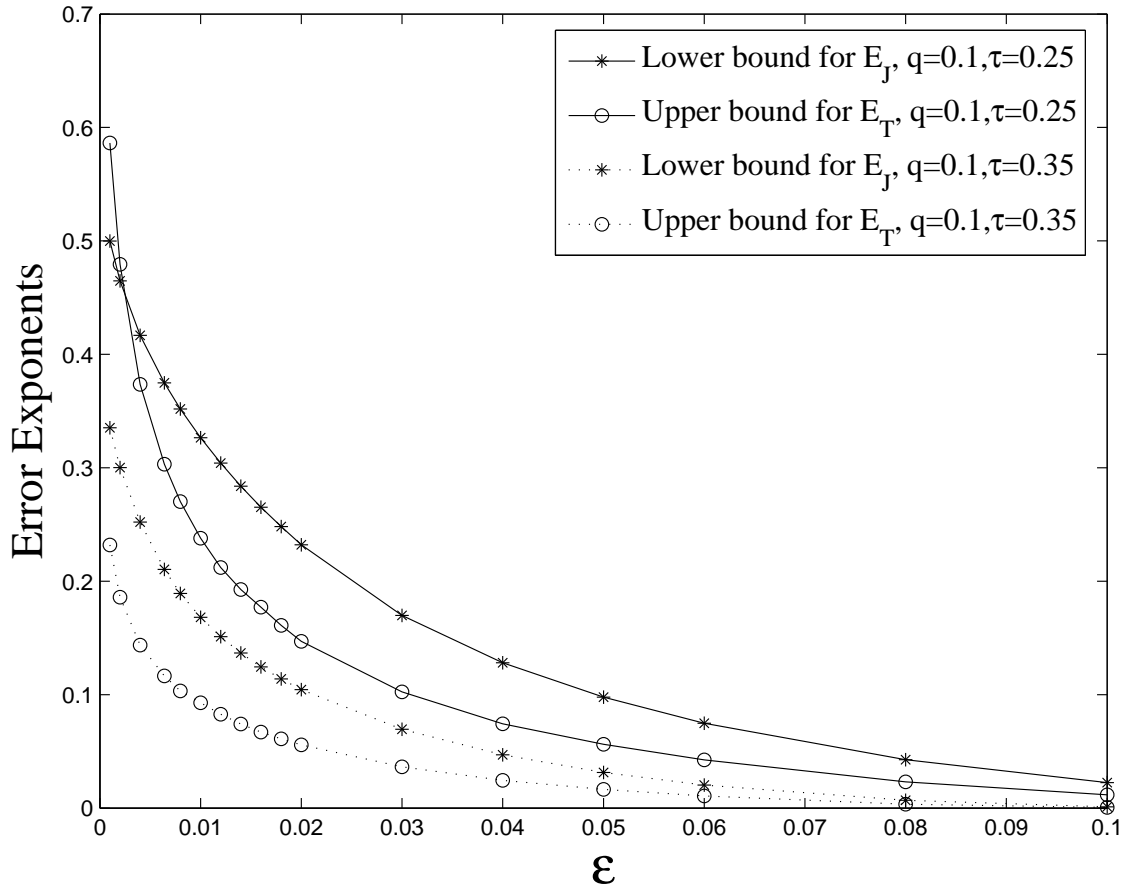


Figure 7: The lower bound of  $E_J$  vs the upper bound of  $E_T$ .