State-Dependent Interference Channel With Correlated States

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Abstract—This paper investigates the Gaussian state-dependent interference channel (IC) and Z-IC, in which two receivers are corrupted respectively by two different but correlated states that are noncausally known to two transmitters but are unknown to the receivers. Three interference regimes are studied, and the capacity region boundary or the sum capacity boundary is characterized either fully or partially under various channel parameters. In particular, the impact of the correlation between states on cancellation of state and interference as well as achievability of capacity is explored with numerical illustrations. For the very strong interference regime, the capacity region is achieved by the scheme where the two transmitters implement a cooperative dirty paper coding. For the strong but not very strong interference regime, the sum-rate capacity is characterized by rate splitting, layered dirty paper coding, and successive cancellation. For the weak interference regime, the sum-rate capacity is achieved via dirty paper coding individually at two receivers as well as treating interference as noise. This paper also provides the achievable region for the state-dependent IC with each state known at its corresponding transmitter.

Index Terms—Capacity region, channel state, dirty paper coding, Gel'fand-Pinsker scheme, noncausal state information, interference channel.

I. INTRODUCTION

TATE-DEPENDENT interference channels (ICs) are of great interest in wireless communications, in which receivers are interfered not only by other transmitters' signals but also by independent and identically distributed (i.i.d.) state sequences. The state can capture interference signals that are informed to transmitters, and are hence often assumed to

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be noncausally known by these transmitters in the model. Such interference cognition can occur in practical wireless networks due to node coordination or backhaul networks. For example, in a cellular network with the coexistence of cellular transmissions and device to device communications defined in release 14 [1] or later networks, and in WLAN standards, researchers are considering multi-access point cooperation in which the knowledge of the channel state information is shared among devices in the next generation WiFi.

State-dependent channels have been widely studied for multiple-access channels (MAC), broadcast channels (BC), and the more complicated interference channels. The two-user MAC with state noncausally known at the transmitters has been studied in various cases. More specifically, [2], [3] studied the model with state noncausally known at both transmitters, while [4] assumed that the state is known only to one transmitter. Reference [5] studied the cognitive model in which one transmitter also knows the other transmitter's message in addition to the noncausal state information. Furthermore, [6], [7] studied the model with the receiver being corrupted by two independent states and each state is known noncausally to one transmitter. For broadcast channel, in [3] and [8]–[11], the state-dependent broadcast channel with private messages and with noncausal state information at the transmitter has been studied. For the state-dependent IC and Z-IC, the state-dependent IC was studied in [12] and [13] with two receivers corrupted by the same state, and in [14] with two receivers corrupted by independent states. In [15] and [16], two state-dependent cognitive IC models were studied, where one transmitter knows both messages, and the two receivers are corrupted by two states. More recently, in [17], both the state-dependent regular IC and Z-IC were studied, where the receivers are corrupted by the same but differently scaled state. In [18] and [19], the interference channel with non-causal state information was studied. Furthermore, in [20] and [21], a type of the state-dependent Z-IC was studied, in which only one receiver is corrupted by the state and the state information is known only to the other transmitter. In [22], a class of deterministic state-dependent Z-ICs was studied, where two receivers are corrupted by the same state and the state information is known only to one transmitter. In [23], a type of the state-dependent Z-IC with two states was studied, where each transmitter knows only the state that corrupts its corresponding receiver. In [24], a state-dependent Z-interference broadcast channel was studied, in which one transmitter has

only one message for its corresponding receiver, and the other transmitter has two messages respectively for two receivers. Both receivers are corrupted by the same state, which is known to both transmitters.

In all the previous work of the state-dependent IC and Z-IC, the states at two receivers are either assumed to be independent, or to be the same but differently scaled, with the exception of [23] that allows correlation between states. However, [23] assumes that each transmitter knows only one state at its corresponding receiver, and hence two transmitters cannot cooperate to cancel the states. In this paper, we investigate the state-dependent IC and Z-IC with the two receivers being corrupted respectively by two correlated states. We studied two cases 1. both transmitters know both states in order for them to cooperate; 2.each transmitter only know the state that is corrupting its targeted receiver. The main focus of this paper is on the Gaussian state-dependent IC and Z-IC, where the receivers are corrupted by additive interference, state, and noise. The aim is to design encoding and decoding schemes to handle interference as well as to cancel the state at the receivers. In particular, we are interested in answering the following two questions: (1) whether and under what conditions both states can be simultaneously fully canceled so that the capacity for the IC and Z-IC without state can be achieved; and (2) what is the impact of the correlation between two states on state cancellation and capacity achievability.

We summarize our results as follows. Our novelty of designing achievable schemes lies in joint design of the interference cancellation schemes together with the Gel'fand-Pinsker binning [25] and dirty paper coding [26] for state cancellation in order to characterize the capacity region. More specifically, we study three interference regimes for the case that both transmitter know both state noncausally. For the very strong interference regime, we characterize the channel parameters under which the two receivers achieve their corresponding point-to-point channel capacity without state and interference. Thus, the interference as well as states are fully canceled, and the capacity region is characterized as a rectangular region. In particular, we demonstrate the impact of the correlation between the two states in such a regime. Interestingly, we show that high interference may not always be beneficial for canceling both state and interference, which is in contrast to the IC without state. For the strong interference regime, we characterize the sum capacity boundary partially under certain channel parameters based on joint design of rate splitting, successive cancellation, as well as dirty paper coding. We also explain how the correlation affects the achievability of the sum capacity, and affects the comparison between the IC and Z-IC. For the weak interference regime, we observe that the sum capacity can be achieved by the two transmitters independently employing dirty paper coding and receiver 1 treating interference as noise as shown in [17] for the same but differently scaled state at the two receivers. The sum capacity is not affected by the correlation between states. We further investigate the state-dependent IC and Z-IC, where each state is known at its corresponding transmitter. We derive the achievable regions for the discrete memoryless and Gaussian channels, where the correlation of the state plays

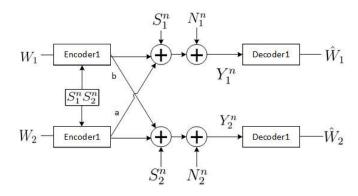


Fig. 1. The state-dependent IC.

an important role. We also show that the sum-capacity can be obtained by treating the other user's state as noise.

II. CHANNEL MODEL

We consider the state-dependent IC (as shown in Fig. 1), in which transmitters 1 and 2 send messages W_1 and W_2 respectively to the receivers 1 and 2. For k = 1, 2, encoder k maps the message $w_k \in \mathcal{W}_k$ to a codeword $x_k^n \in \mathcal{X}_k^n$. The two inputs x_1^n and x_2^n are then transmitted over the IC to the receivers, which are corrupted by two correlated state sequences S_1^n and S_2^n , respectively. We considered two cases for the channel model, i.e., 1. the state sequences are known to both the transmitters noncausally, but are unknown at the receivers; 2. the state sequence is known to its corresponding transmitter. Encoders 1 and 2 want to map their messages as well as the state sequence(s)' information into codewords $x_1^n \in \mathcal{X}_1^n$ and $x_2^n \in \mathcal{X}_2^n$. The channel transition probability is given by $P_{Y_1Y_2|S_1S_2X_1X_2}$. The decoders at the receivers map the received sequences y_1^n and y_2^n into corresponding messages $\hat{w}_k \in \mathcal{W}_k$ for k = 1, 2.

The average probability of error for a length-n code is defined as

$$P_e^{(n)} = \frac{1}{|\mathcal{W}_1||\mathcal{W}_2|} \sum_{w_1=1}^{|\mathcal{W}_1|} \sum_{w_2=1}^{|\mathcal{W}_2|} Pr\{(\hat{w}_1, \hat{w}_2) \neq (w_1, w_2)\}.$$
(1)

A rate pair (R_1, R_2) is *achievable* if there exist a sequence of message sets $\mathcal{W}_k^{(n)}$ with $|\mathcal{W}_k^{(n)}| = 2^{nR_k}$ for k = 1, 2, such that the average error probability $P_e^{(n)} \to 0$ as $n \to \infty$. The *capacity region* is defined to be the closure of the set of all achievable rate pairs (R_1, R_2) .

In this paper, we study the Gaussian channel with the outputs at the two receivers for one channel use given by

$$Y_1 = X_1 + aX_2 + S_1 + N_1, (2a)$$

$$Y_2 = bX_1 + X_2 + S_2 + N_2 \tag{2b}$$

where a and b are the channel gain coefficients, and N_1 and N_2 are noise variables with Gaussian distributions $N_1 \sim \mathcal{N}(0, 1)$ and $N_2 \sim \mathcal{N}(0, 1)$. The state variables S_1 and S_2 are jointly Gaussian with the correlation coefficient ρ and the marginal distributions $S_1 \sim \mathcal{N}(0, Q_1)$ and $S_2 \sim \mathcal{N}(0, Q_2)$. Both the noise variables and the state variables are i.i.d. over the

channel uses. The channel inputs X_1 and X_2 are subject to the average power constraints P_1 and P_2 .

The Z-IC is also studied in this paper, in which receiver 2 is not corrupted by the interference X_1 from transmitter 1 (i.e, b = 0 for Gaussian channel).

Our goal is to investigate state and interference cancellation scheme and characterize the capacity region for the state-dependent IC and Z-IC, which matches the capacity of the corresponding IC and Z-IC without state corruption. Moreover, we identify channel parameters, under which the capacity region of the state-dependent IC and Z-IC can be established. In particular, we are interested in understanding the impact of the correlation between the states S_1 and S_2 on the capacity characterization. We will also investigate the case when each transmitter only knows its corresponding channel state.

III. VERY STRONG INTERFERENCE REGIME

In this section, we study the impact of the correlation between states on the characterization of the capacity in the very strong interference regime. We study both the state-dependent IC and Z-IC.

A. State-Dependent IC

In this subsection, we study the state-dependent IC with channel states known to both transmitters in the very strong interference regime, where the channel parameters satisfy

$$P_1 + a^2 P_2 + 1 > 1 + P_1)(1 + P_2),$$
 (3a)

$$b^2 P_1 + P_2 + 1 > (1 + P_1)(1 + P_2).$$
 (3b)

For the corresponding IC without states, the capacity region contains rate pairs (R_1, R_2) satisfying:

$$R_1 \leqslant \frac{1}{2} \log(1 + P_1),$$

 $R_2 \leqslant \frac{1}{2} \log(1 + P_2).$ (4)

In this case, the two receivers achieve the point-to-point channel capacity without interference. Furthermore, in [17], an achievable scheme has been established to achieve the same point-to-point channel capacity when the two receivers are corrupted by the same but differently scaled state. Our focus here is on the more general scenario, where the two receivers are corrupted by two *correlated* states, and our aim is to understand how the correlation affects the design of the scheme.

We first design an achievable scheme to obtain an achievable rate region for the discrete memoryless IC. The two transmitters encode their messages W_1 and W_2 into two auxiliary random variables U and V, respectively, based on the Gel'fand-Pinsker binning scheme. Since the channel satisfies the very strong interference condition, it is easier for receivers to decode the information of the interference. Thus each receiver first decodes the auxiliary random variable corresponding to the message intended for the other receiver, and then decodes its own message by decoding the auxiliary random variable for itself. For instance, receiver 1 first

decodes V, then uses it to cancel the interference X_2 and partial state interference, and finally decodes its own message W_1 by decoding U. Differently from [17], two auxiliary random variables U and V are designed not with regard to one state, but with regard to two correlated states. This requires a joint design for U and V to fully cancel the states. Based on such a scheme, we obtain the following achievable region.

Proposition 1. For the state-dependent IC with states non-causally known at both transmitters, the achievable region consists of rate pairs (R_1, R_2) satisfying:

$$R_1 \leq \min\{I(U; VY_1), I(U; Y_2)\} - I(S_1S_2; U),$$
 (5a)

$$R_2 \leq \min\{I(V; UY_2), I(V; Y_1)\} - I(S_1S_2; V)$$
 (5b)

for some distribution $P_{S_1S_2}P_{U|S_1S_2}P_{X_1|US_1S_2}P_{V|S_1S_2}$ $P_{X_2|VS_1S_2}P_{Y_1Y_2|S_1S_2X_1X_2}$, where U and V are auxiliary random variables.

Proof: See Appendix A.
$$\Box$$

We now study the Gaussian IC. For the sake of technical convenience, we express the Gaussian channel in Section II in a different form. Since S_1 and S_2 are jointly Gaussian, S_1 can be expressed as $S_1 = dS_2 + S_1'$ where d is a constant representing the level of correlation, and S_1' is independent from S_2 and $S_1' \sim \mathcal{N}(0, Q_1')$ with $Q_1 = d^2Q_2 + Q_1'$. Thus, without loss of generality, the channel model can be expressed in the following equivalent form that is more convenient for analysis.

$$Y_1 = X_1 + aX_2 + dS_2 + S_1' + N_1, (6a)$$

$$Y_2 = bX_1 + X_2 + S_2 + N_2. (6b)$$

Following Proposition 1, we characterize the condition under which both the state and interference can be fully canceled, and hence the capacity region for the state-dependent Gaussian IC in the very strong interference regime is obtained.

Theorem 1. For the state-dependent Gaussian IC with state noncausally known at both transmitters, the capacity region is the same as the point-to-point channel capacity for both receivers, if the channel parameters $(a, b, d, P_1, P_2, Q'_1, Q_2)$ satisfy the following conditions:

$$\frac{1}{2}\log(1+P_1) \leqslant h(X_1) - h(U,Y_2) + h(Y_2), \tag{7a}$$

$$\frac{1}{2}\log(1+P_2) \leqslant h(X_2) - h(V,Y_1) + h(Y_1), \tag{7b}$$

where the auxiliary random variables are designed as $U = X_1 + \alpha_1 S_1' + \alpha_2 S_2$ and $V = X_2 + \beta_1 S_1' + \beta_2 S_2$. Here, X_1, X_2, S_1' and S_2 are independent Gaussian variables with mean zero and variances P_1, P_2, Q_1 and Q_2 , respectively. The parameters $\alpha_1, \alpha_2, \beta_1$ and β_2 are set as

$$\alpha_{1} = \frac{P_{1}(1+P_{2})}{(P_{1}+1)(P_{2}+1) - abP_{1}P_{2}},$$

$$\alpha_{2} = \frac{P_{1}(d+dP_{2}-aP_{2})}{(P_{1}+1)(P_{2}+1) - abP_{1}P_{2}},$$

$$\beta_{1} = \frac{bP_{1}P_{2}}{(P_{1}+1)(P_{2}+1) - abP_{1}P_{2}},$$

$$\beta_{2} = \frac{P_{2}(P_{1}+1 - bdP_{1})}{(P_{1}+1)(P_{2}+1) - abP_{1}P_{2}}.$$
(8)

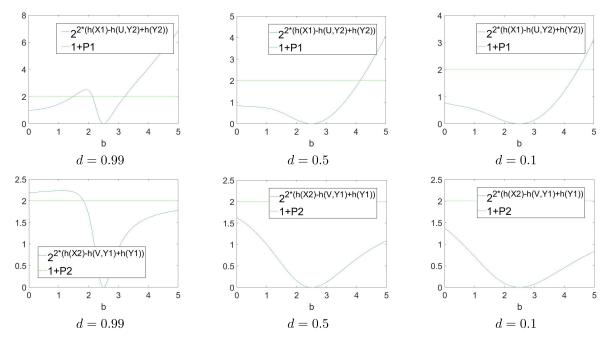


Fig. 2. Comparison of values of both sides of (13a) and (13b).

Proof: The proof mainly follows Proposition 1. We design U and V as stated in Theorem 1. As discussed in the proof of Proposition 1, V is first decoded by receiver 1 and U is first decoded by receiver 2. And then receiver 2 subtracts Uto cancel X_1 and obtain $Y_2' = Y_2 - bU = X_2 - b\alpha_1 S_1' + (1 - b\alpha_1 S_1' + 1)$ $b\alpha_2)S_2 + N_2$, and receiver 1 subtracts V to cancel X_2 and obtain $Y_1' = Y_1 - aV = X_1 + (1 - a\beta_1)S_1' + (d - a\beta_2)S_2 + N_1$. In order to fully cancel the channel states for Y_1' and Y_2' , based on the dirty paper coding scheme, we further require the coefficients to satisfy the following conditions,

$$\frac{\alpha_1}{1 - a\beta_1} = \frac{\alpha_2}{d - a\beta_2} \tag{9a}$$

$$\frac{a_1}{1 - a\beta_1} = \frac{P_1}{P_1 + 1} \tag{9b}$$

$$\frac{\beta_1}{-b\alpha_1} = \frac{\beta_2}{1 - b\alpha_2} \tag{9c}$$

$$\frac{\beta_1}{-b\alpha_1} = \frac{P_2}{P_2 + 1} \tag{9d}$$

which yield $\alpha_1, \alpha_2, \beta_1$ and β_2 in (8).

By plugging these parameters into (5a), we obtain

$$I(U; VY_1) - I(S_1, S_2; U) = \frac{1}{2} \log(1 + P_1),$$

which yields

$$R_1 \leqslant \min\{I(U; Y_2) - I(S_1, S_2; U), \frac{1}{2}\log(1 + P_1)\}.$$
 (10)

Similarly, (5b) yields

$$R_2 \leq \min\{I(V; Y_1) - I(S_1, S_2; V), \frac{1}{2}\log(1 + P_2)\}.$$
 (11)

In order to achieve the channel capacity of the point-to-point channel as shown in (4) for both receivers, the following

conditions should be satisfied:

$$\frac{1}{2}\log(1+P_1) \leqslant I(U; Y_2) - I(S_1, S_2; U) \tag{12a}$$

$$\frac{1}{2}\log(1+P_2) \leqslant I(V; Y_1) - I(S_1, S_2; V), \tag{12b}$$

$$\frac{1}{2}\log(1+P_2) \leqslant I(V;Y_1) - I(S_1, S_2; V), \quad (12b)$$

which completes the proof.

We note that the conditions egrefeq:cond1 and (7b) in Theorem 1 is the very strong condition for the interference channel with state and represent the comparison between the ability of receivers to decode messages in different decoding steps. For instance, in condition (7a) the right-hand side term represents how much receiver 2 can decode U in the first step of decoding in order to cancel the interference, and the left-hand side term represents the rate at which receiver 1 can decode U in the second step of decoding, where we can use the dirty paper coding scheme to fully cancel the states and achieve the capacity. Hence, achieving the point-to-point channel capacity requires the second step to dominate the performance.

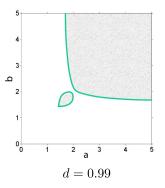
We next study the impact of the channel parameters and state correlation on the achievablility of the point-to-point capacity. In particular, we illustrate how the interference gains (a, b) affect the conditions (7a) and (7b). To make the figure more clear, we take exponential of both sides, and hence the conditions (7a) and (7b) become:

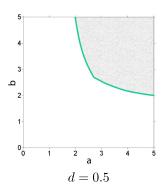
$$1 + P_1 \leqslant 2^{2(h(X_1) - h(U, Y_2) + h(Y_2))}, \tag{13a}$$

$$1 + P_1 \leqslant 2$$

$$1 + P_2 \leqslant 2^{2(h(X_2) - h(V, Y_1) + h(Y_1))}.$$
(13b)

In Fig. 2, we set $Q_1 = Q_2 = 0.9, P_1 = 1, P_2 = 1$ and a = 1.6, and plot the change of both left-hand side terms and right-hand side terms in (13a) and (13b) versus the channel parameters b for three different values of d. Taking the first row of Fig. 2 as an example, it is clear that





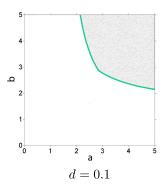


Fig. 3. Ranges of (a,b) under which the point-to-point channel capacity is achieved

 $1 + P_1$ is a straight line, and $2^{2(h(X_1) - h(U,Y_2) + h(Y_2))}$ is not a monotone function with respect to b. The condition (13a) (and hence (7a)) is satisfied only when $2^{2(h(X_1) - h(U,Y_2) + h(Y_2))}$ is above the straight line $1 + P_1$. When the parameter d = 0.99, there are two regions over which the condition (13a) is satisfied. But if d = 0.5, there is only one region over which the condition (13a) is satisfied. For d = 0.1, there is also only one region where the condition (13a) is satisfied. Similarly, the second row in Fig. 2 illustrates the regions of b over which the condition (13b) is satisfied for the corresponding values of d. Then the intersection of the region of b in the first and second rows of Fig. 2 fully determines the ranges of b over which the point-to-point channel capacity can be achieved for both receivers.

The range of the parameters (a,b) such that the point-to-point channel capacity is obtained is shown in Fig. 3. For these figures, if we fix a=1.6, then the ranges of b is consistent with those in Fig. 3 where both (13a) and (13b) are satisfied. Fig. 3 also illustrates the impact of the correlation d between the states S_1 and S_2 on the achievability of channel capacity. It is clear that as d increases, i.e., the two states are more correlated, the range of (a,b) over which the point-to-point channel capacity is achieved gets larger. This confirms the intuition that more correlated states are easier to be fully canceled.

B. State-Dependent Z-IC

In this subsection, we study the state-dependent Z-IC in the very strong interference regime, in which the channel parameters satisfy $a^2 > 1 + P_1$. Similarly to the state-dependent IC, we study the ranges of the channel parameters, over which the capacity is also characterized by (4). We note that here Z-IC (with b = 0) cannot be viewed as a special case of the IC in the very strong interference regime.

We first design an achievable scheme to obtain an achievable rate region for the discrete memoryless Z-IC. The two transmitters encode their messages W_1 and W_2 into two auxiliary random variables U and V, respectively, based on the Gel'fand-Pinsker binning scheme. Since receiver 2 is interference free and is corrupted by S_2 , the auxiliary random variable V is designed with regard to only S_2 . Furthermore, receiver 1 first decodes V, then uses it to cancel the interference X_2 and partial state interference, and finally decodes

its own message W_1 by decoding U. Here, since S_2 is introduced to Y_1 when canceling X_2 via V, the auxiliary random variable U is designed based on both S_1 and S_2 to fully cancel the states. Based on such a scheme, we obtain the following achievable region.

Proposition 2. For the state-dependent Z-IC with the states noncausally known at both transmitters, an achievable region consists of rate pairs (R_1, R_2) satisfying:

$$R_1 \leqslant I(U; VY_1) - I(S_1S_2; U)$$
 (14a)

$$R_2 \leq \min\{I(V; Y_2), I(V; Y_1)\} - I(S_2; V)$$
 (14b)

for some distribution $P_{S_1S_2}P_{U|S_1S_2}P_{X_1|US_1S_2}P_{V|S_2}P_{X_2|VS_2}$ $P_{Y_1|S_1X_1X_2}P_{Y_2|S_2X_2}$.

Proof: See Appendix B.
$$\Box$$

Following Proposition 2, we further simplify the achievable region in the following corollary, which is in a useful form for us to characterize the capacity region for the Gaussian Z-IC.

Corollary 1. For the state-dependent Z-IC with the states noncausally known at both transmitters, if the following condition

$$I(V; Y_2) \leqslant I(V; Y_1) \tag{15}$$

is satisfied, then an achievable region consists of rate pairs (R_1, R_2) satisfying:

$$R_1 \le I(U; VY_1) - I(S_1S_2; U)$$

 $R_2 \le I(V; Y_2) - I(S_2; V)$ (16)

for some distribution $P_{S_1S_2}P_{U|S_1S_2}P_{X_1|US_1S_2}P_{V|S_2}P_{X_2|VS_2}$ $P_{Y_1|S_1X_1X_2}P_{Y_2|S_2X_2}$.

In Corollary 1, condition (15) requires that receiver 1 is more capable in decoding V (and hence W_2) than receiver 2, which is likely to be satisfied in the very strong interference regime.

Following Corollary 1, we characterize the channel parameters under which both the states and interference can be fully canceled, and hence the capacity region for the Z-IC is obtained.

Theorem 2. For the state-dependent Gaussian Z-IC with states noncausally known at both transmitters, if the channel parameters $(a, d, P_1, P_2, Q'_1, Q_2)$ satisfy the following

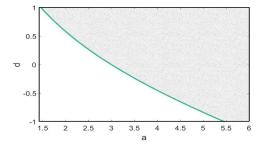


Fig. 4. Characterization of channel parameters (a,d) in shaded area under which the state-dependent Gaussian Z-IC achieves the capacity of the corresponding channel without states and interference in very strong interference regime.

condition:

$$\frac{P_1 + a^2 P_2 + d^2 Q_2 + Q_1' + 1}{(d + a\beta)^2 Q_2 P_2 + (P_2 + \beta^2 Q_2)(P_1 + Q_1' + 1])} \geqslant \frac{P_2 + 1}{P_2}$$
(17)

where $\beta = \frac{P_2}{P_2+1}$, then the capacity region is characterized by (4).

Proof: Theorem 2 follows from Corollary 1 by setting $U = X_1 + \alpha_1 S_2 + \alpha_2 S_1'$, $V = X_2 + \beta S_2$, where X_1 , X_2 , S_1' and S_2 are independent Gaussian variables withe mean zero and variances P_1 , P_2 , Q_1' and Q_2 , respectively. As discussed in the proof of Proposition 2, V is first decoded by decoder 1. And then by dirty paper coding, we design α_1 , α_2 and β for both $Y_2 = X_2 + S_2 + N_2$ and $Y_1' = Y_1 - aV = X_1 + (d - a\beta)$ $S_2 + S_1' + N_1$ to fully cancel the states. Thus, the coefficients should satisfy the following conditions:

$$\frac{\alpha_1}{d - a\beta} = \frac{P_1}{P_1 + 1} \tag{18a}$$

$$\alpha_2 = \frac{P_1}{P_1 + 1} \tag{18b}$$

$$\beta = \frac{P_2}{P_2 + 1},\tag{18c}$$

which further yields α_1 , α_2 and β that satisfy

$$\alpha_1 = \frac{P_1}{P_1 + 1}(d - \frac{aP_2}{P_2 + 1}), \quad \alpha_2 = \frac{P_1}{P_1 + 1}, \quad \beta = \frac{P_2}{P_2 + 1}.$$

Substituting the above choice of the auxiliary random variables and the parameters into (15) in Corollary 1, we obtain the condition (17). Substituting those choices into the condition (16), we obtain the capacity region characterized by (4). Since such an achievable region achieves the point-to-point channel capacity for the Z-IC without the state, it can be shown to be the capacity region of the state-dependent Z-IC.

Based on Theorem 2, if channel parameters satisfy the condition (17), we can simultaneously cancel two states and the interference, and the point-to-point capacity of two receivers without state and interference can be achieved. The correlation between the two states captured by d plays a very important role regarding whether the condition can be satisfied. In Fig. 4, we set $P_1 = 2$, $P_2 = 2$, $Q_1 = 1$ and $Q_2 = 1$, and plot the range of the parameter pairs (a, d) under which the channel capacity without states and interference can be achieved.

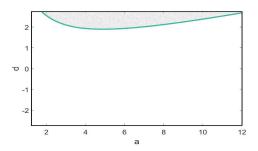


Fig. 5. Characterization of channel parameters (a, d) in shaded area under which the state-dependent Gaussian Z-IC achieves the capacity of the corresponding channel without states in very strong interference regime when $Q_2 > \frac{1+P_2}{P_0}$.

These parameters fall in the shaded area above the line. It can be seen that as d becomes larger (i.e., the correlation between the two states increases), the threshold on the parameter a to fully cancel the interference and state becomes smaller. This suggests that more correlated states are easier to cancel together with the interference.

Fig. 4 agrees with the result of the very strong IC without states in the sense that once a is above a certain threshold (i.e., the interference is strong enough), then the point-topoint channel capacity without interference can be achieved. However, this is not always true for the *state-dependent* Z-IC. This can be seen from the condition (17) in Theorem 2. If we let a go to infinity, then the condition (17) becomes $Q_2 > \frac{1+P_2}{P_2}$, which is not always satisfied. This is because in the presence of state, Y_1 decodes V instead of X_2 , and the decoding rate is largest if the dirty paper coding design of V (based on S_2 at receiver 2) also happens to be the same dirty paper coding design against S_2 at receiver 1. Clearly, as a gets too large, V is more deviated from such a favorable design, and hence the decoding rate becomes smaller, which consequently hurts the achievability of the point-to-point capacity for receiver 2. Such a phenomena can be observed in Fig. 5, where the constant a cannot be too large to guarantee the achievability of the point-to-point channel capacity. Furthermore, the figure also suggests that further correlated states allow a larger range of a under which the point-to-point channel capacity can be achieved.

IV. STRONG INTERFERENCE REGIME

In this section, we study the state dependent IC in the strong interference regime, which excludes the very strong interference regime that has been studied in Section III. With the presence of states, these two regimes require separate treatments for state cancellation.

A. State-Dependent Regular IC

For the corresponding IC without state, if it is strong but not very strong, then the channel parameters satisfy

$$a \ge 1$$
, $b \ge 1$, (19)
 $\min\{P_1 + a^2P_2 + 1, b^2P_1 + P_2 + 1\} \le (1 + P_1)(1 + P_2)$.

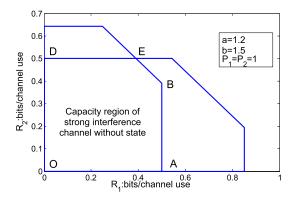


Fig. 6. Capacity region of the strong IC without state

Without loss of generality, we assume that $P_1 + a^2P_2 + 1 \le b^2P_1 + P_2 + 1$. It has been shown in [27] that the capacity region for the strong IC without states contains rate pair (R_1, R_2) satisfying

$$R_1 \le \frac{1}{2}\log(1+P_1), \quad R_2 \le \frac{1}{2}\log(1+P_2),$$

 $R_1 + R_2 \le \frac{1}{2}\log(P_1 + a^2P_2 + 1).$ (20)

Such a region is an intersection of the capacity regions of two MACs, which is illustrated as the pentagon O-A-B-E-D-O in Fig. 6. Our goal here is to study whether the points on the sum-rate capacity boundary of the IC without state can be achieved with the presence of state. Such a problem has been studied in [17] for the channel with two receivers corrupted by the same but differently scaled state. Here, we generalize such a study to the situation when the two receivers are corrupted by two correlated states.

Since every point on this line of the sum-rate capacity can be achieved by rate splitting and successive cancellation in the case without state, for the state-dependent channel, we continue to adopt the idea of rate splitting and successive cancellation but using auxiliary random variables to incorporate dirty paper coding to further cancel state successively. More specifically, transmitter 1 splits its message W_1 into W_{11} and W_{12} , and then encodes them into U_1 and U_2 respectively based on the Gel'fand-Pinsker binning scheme. Then transmitter 2 encodes its message W_2 into V, based on the Gel'fand-Pinsker binning scheme. The auxiliary random variables U_1 , U_2 , and V are designed such that decoding of them at receiver 1 successively fully cancels the state corruption of Y_1 so that the sum capacity boundary (i.e., the line B-E) can be achieved if only decoding at receiver 1 is considered. Now further incorporating the decoding at receiver 2, if for any point on the line B-E, decoding of V at receiver 2 does not cause further rate constraints, then such a point is achievable for the state-dependent IC. Here, it is clear that due to the presence of the states at both receivers as well as our design of the binning scheme to fully cancel the state corruption at receiver 1, there is a necessary change in receiver 2' decoding capability. In particular, the corner point E (which is achievable for the case without state) is not achievable any more.

Proposition 3. For the state-dependent IC with states non-causally known at both transmitters, an achievable region consists of rate pairs (R_1, R_2) satisfying:

$$R_{1} \leqslant \min\{I(U_{1}; Y_{1}), I(U_{1}; Y_{2})\}$$

$$+ \min\{I(U_{2}; VY_{1}|U_{1}), I(U_{2}; VY_{2}|U_{1})\} - I(U_{1}U_{2}; S_{1})$$

$$R_{2} \leqslant \min\{I(V; Y_{1}|U_{1}), I(V; Y_{2}|U_{1})\} - I(V; S_{1})$$
(21)

for some distribution $P_{S_1S_2}P_{V|S_1}P_{X_2|VS_1}P_{U_1|S_1}P_{U_2|S_1U_1}P_{X_1|S_1U_1U_2}P_{Y_1|S_1X_1X_2}P_{Y_2|S_2X_2}$, where U_1 , U_2 and V are auxiliary random variables.

Proof: See Appendix C.
$$\Box$$

Remark 1. This scheme can be generalized through further splitting the messages and changing the orders of decoding the messages at the two receivers. The achievable region can then be obtained by taking the convex hull of the union over all achievable regions corresponding to different schemes above.

Based on Proposition 3, we next characterize partial boundary of the capacity region for the state-dependent Gaussian IC. For the sake of technical convenience, we express the Gaussian model in a different form. In particular, we express S_2 as $S_2 = cS_1 + S_2'$ where c is a constant representing the level of correlation, and S_1 is independent from S_2' with $S_2' \sim \mathcal{N}(0, Q_2')$ where $Q_2 = c^2Q_1 + Q_2'$. Thus, without loss of generality, the channel model can be expressed in the following equivalent form that is more convenient for analysis here.

$$Y_1 = X_1 + aX_2 + S_1 + N_1 \tag{22a}$$

$$Y_2 = bX_1 + X_2 + cS_1 + S_2' + N_2.$$
 (22b)

We next show that we can design a scheme to achieve the partial boundary of the capacity region for the IC without state. We note that the rate on the sum-capacity boundary can be characterized by

$$R_{1} = \frac{1}{2} \log \left(1 + \frac{P_{1}'}{a^{2} P_{2} + P_{1}'' + 1} \right) + \frac{1}{2} \log \left(1 + P_{1}'' \right),$$

$$R_{2} = \frac{1}{2} \log \left(1 + \frac{a^{2} P_{2}}{P_{1}'' + 1} \right),$$
(23)

for some $P'_1, P''_1 \ge 0$, and $P'_1 + P''_1 \le P_1$.

Theorem 3. Any rate point in (23) can be achieved by the state-dependent IC if the channel parameters satisfy the following conditions

$$I(U_1; Y_2) - I(U_1; S_1) \leq \frac{1}{2} \log \left(1 + \frac{P_1'}{a^2 P_2 + P_1'' + 1} \right)$$

$$I(U_2; V Y_2 | U_1) - I(U_2; S_1 | U_1) \leq \frac{1}{2} \log \left(1 + P_1'' \right)$$

$$I(V; Y_2 | U_1) - I(V; S_1) \leq \frac{1}{2} \log \left(1 + \frac{a^2 P_2}{P_1'' + 1} \right), \quad (24)$$

where the mutual information terms are calculated by setting $U_1 = X_1' + \alpha_1 S_1$, $U_2 = X_1'' + \alpha_2 S_1$, $V = a X_2 + \beta S_1$ and $X_1 = X_1' + X_1''$. Here X_1' , X_1'' and X_2 are Gaussian variables with mean zero and variances P_1' , P_1'' and P_2 , and P_3 , P_4 and P_4 and P_4 and P_4 and P_4 and P_4 are

and β are given by

$$\alpha_1 = \frac{P_1'}{P_1 + a^2 P_2 + 1}, \quad \alpha_2 = \frac{P_1''}{P_1 + a^2 P_2 + 1},$$

$$\beta = \frac{a^2 P_2}{P_1 + a^2 P_2 + 1}.$$
(25)

Proof: Theorem 3 follows from Proposition 3 by choosing the auxiliary random variables U_1 , U_2 and V as in the statement of the theorem. In particular, U_1 is first decoded by receiver 1, and is designed to cancel the state in Y_1 treating all other variables as noise. Then, V is decoded by receiver 1, and is designed to cancel the state in $Y_1' = Y_1 - U_1 = X_1'' + aX_2 + (c - \alpha_1)S_1 + N_1$. Finally, U_2 is designed to cancel the state in $Y_1'' = Y_1' - V = X_1'' + (c - \alpha_1 - \beta)S_1 + N_1$. In order to satisfy the state cancellation requirements, α_1 , α_2 and β should satisfy

$$\alpha_1 = \frac{P_1'}{P_1 + a^2 P_2 + 1},\tag{26}$$

$$\frac{\alpha_2}{1 - \alpha_1} = \frac{P_1''}{P_1'' + 1},\tag{27}$$

$$\frac{\beta}{1-\alpha_1} = \frac{a^2 P_2}{P_1'' + a^2 P_2 + 1},\tag{28}$$

which yields (25). Substituting these choices of the random variables and the coefficients into Proposition 3, (21) becomes (29).

$$R_{1} \leq \min \left\{ I(U_{1}; Y_{2}) - I(U_{1}; S_{1}), \frac{1}{2} \log \left(1 + \frac{P_{1}'}{a^{2}P_{2} + P_{1}'' + 1} \right) \right\}$$

$$+ \min \left\{ I(U_{2}; VY_{2}|U_{1}) - I(U_{2}; S_{1}|U_{1}), \frac{1}{2} \log \left(1 + P_{1}'' \right) \right\}$$

$$R_{2} \leq \min \left\{ I(V; Y_{2}|U_{1}) - I(V; S_{1}), \frac{1}{2} \log \left(1 + \frac{a^{2}P_{2}}{P_{1}'' + 1} \right) \right\}.$$

$$(29)$$

Hence, if the condition (24) is satisfied, the points on the sum capacity boundary (23) can be achieved.

B. State-Dependent Z-IC

In this subsection, we study the state-dependent Z-IC in the strong, but not very strong interference regime, in which the channel parameters satisfy $1 \leqslant a^2 < 1 + P_1$. For the corresponding Z-IC without states, it has been shown that the capacity region contains rate pairs (R_1, R_2) satisfying

$$R_1 + R_2 \leqslant \frac{1}{2} \log(1 + P_1 + a^2 P_2)$$

 $R_1 \leqslant \frac{1}{2} \log(1 + P_1), \quad R_2 \leqslant \frac{1}{2} \log(1 + P_2)$ (30)

which is illustrated as the pentagon O-A-B-E-F in Fig. 7.

Our goal here is to study whether the points on the sum-capacity boundary of the Z-IC *without* state (i.e., the line B-E in Fig. 7) can be achieved for the corresponding *state-dependent* Z-IC. We follow the idea for the state-dependent IC to design an achievable scheme, and obtain the following Proposition.

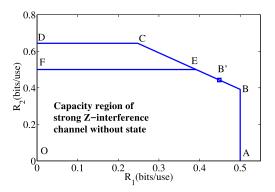


Fig. 7. Capacity region of the strong Z-IC without state.

Proposition 4. For the state-dependent Z-IC with states non-causally known at both transmitters, if the following condition is satisfied

$$I(V; U_1Y_1) \leqslant I(V; Y_2),$$
 (31)

then an achievable region consists of rate pairs (R_1, R_2) satisfying:

$$R_1 \leqslant I(U_1; Y_1) + I(U_2; VY_1|U_1) - I(S_1; U_1U_2)$$
 (32a)

$$R_2 \leqslant I(V; U_1Y_1) - I(S_1; V)$$
 (32b)

 $\begin{array}{lll} for & some & distribution & P_{S_1S_2}P_{V|S_1}P_{X_2|VS_1}P_{U_1|S_1}P_{U_2|S_1U_1} \\ P_{X_1|S_1U_1U_2}P_{Y_1|S_1X_1X_2}P_{Y_2|S_2X_2}. \end{array}$

The proof of Proposition 4 is similar to the proof of Proposition 3. The only difference lies in that Y_2 does not need to decode U_1 and U_2 because of no interference at receiver 2.

For the Gaussian model, based on Proposition 4, we characterize the condition under which any point on the sum capacity boundary of the strong Z-IC without states (e.g., point B' in Fig. 7) is achievable. Hence, such a point is on the sum capacity boundary of the state-dependent Z-IC.

Theorem 4. For the state-dependent Gaussian Z-IC with state noncausally known at both transmitters, if the channel parameters $(a, c, P_1, P_2, Q_1, Q'_2)$ satisfy the following condition:

$$\frac{a^2 P_2 (P_2 + c^2 Q_1 + Q_2' + 1)}{(ac - \beta)^2 Q_1 P_2 + (a^2 P_2 + \beta^2 Q_1)(Q_2' + 1)} \ge 1 + \frac{a^2 P_2}{P_1'' + 1}$$
(33)

where $\beta = \frac{a^2 P_2}{P_1 + a^2 P_2 + 1}$, then the following point (on the line B-E)

$$R_{1} = \frac{1}{2} \log \left(1 + \frac{P_{1}'}{a^{2}P_{2} + P_{1}'' + 1} \right) + \frac{1}{2} \log \left(1 + P_{1}'' \right)$$

$$R_{2} = \frac{1}{2} \log \left(1 + \frac{a^{2}P_{2}}{P_{1}'' + 1} \right)$$
(34)

where $P'_1 = P_1 - P''_1$, is on the sum capacity boundary.

Proof: Theorem 4 follows from Proposition 4 by choosing the auxiliary random variables U_1 , U_2 and V based on the setting in Theorem 3. Thus, the state at Y_1 can be fully canceled. Furthermore, by substituting the auxiliary random variables in Theorem 3 into (31), the condition (33) can be

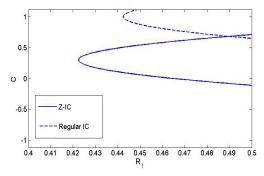


Fig. 8. Ranges of *c* under which points on sum capacity boundary of the strong IC and Z-IC without state can be achieved by the state-dependent IC and Z-IC

obtained, under which the points characterized in (34) can be achieved.

Theorem 4 provides the condition of channel parameters under which a certain given point is on the sum capacity boundary of the capacity region. We next characterize a line segment on the sum capacity boundary for a given set of channel parameters.

Corollary 2. For the state-dependent Z-IC with state non-causally known at both transmitters, if a point on the line B-E in Fig. 7 is on the sum-capacity boundary for a given set of channel parameters, then the segment between this point and point B on the line B-E is on the sum capacity boundary for the same set of channel parameters.

In order to numerically illustrate Theorem 4, we first note that each point on the sum-capacity boundary (i.e., the line B-E in Fig. 7) can be expressed as (R_1, R_2) = $(R_1, \frac{1}{2}\log(P_1 + a^2P_2 + 1) - R_1)$. We now set $P_1 = 2$, $P_2 = 0.7$, $Q_1 = 0.4$, $Q_2 = 0.5$ and a = 1.2, and hence $R_1 \in \left[\frac{1}{2}\log(1.23), 0.5\right]$ parameterizes all points from point E to point B in Fig. 7. In Fig. 8, we plot the ranges of c under which points, parameterized by R_1 on the sum capacity boundary of the strong Z-IC without state, can be achieved by the state-dependent Z-IC following Theorem 4. It can be seen that as correlation between the two states (represented by c) increases, initially more points on the sum capacity boundary are achieved and then less points are achieved as c is above a certain threshold. Thus, higher correlation does not guarantee more capability of achieving the sum capacity boundary. This is because in our scheme both U_i and V are specially designed for Y_1 based on dirty paper coding. At receiver 2, such design of V initially approximates better the dirty paper coding design for Y_2 as c becomes large, but then becomes worse as c continues to increase. Hence decoding of V at receiver 2 initially gets better and then becomes less capable, which consequently determines variation of achievability of the sum capacity boundary.

Fig. 8 also plots the same parameter range for the state-dependent IC as characterized by Theorem 3. It is clear that the state-dependent IC achieves a smaller line segment on the sum-capacity (i.e., smaller range of R_1). This is reasonable, because Theorem 3 for the IC requires more conditions than Theorem 4 for the Z-IC. Fig. 8 also demonstrates that large value of c(i.e., higher correlation between the states) is required for the IC to achieve the sum capacity than the Z-IC.

This is because the dirty paper coding is designed with respect to receiver 1. High correlation between states helps such design to be more effective to cancel state at receiver 2 as well.

V. WEAK INTERFERENCE REGIME

In this section, we study the state-dependent IC and Z-IC in the weak interference regime. The channel parameters for the IC in this regime satisfy $|a(1+b^2P_1)|+|b(1+a^2P_2)|\leqslant 1$, which reduces to $a\leqslant 1$ for the Z-IC. It has been shown in [28]–[30], for the weak IC without state and in [31] for the weak Z-IC that the sum capacity can be achieved by treating interference as noise. It was further shown in [17] that for the IC and Z-IC with the same but differently scaled state at two receivers, independent dirty paper coding at the two transmitters to cancel the states and treating interference as noise achieve the same sum capacity. We here observe that such a scheme is also achievable with the presence of two correlated states, which thus yields the following Corollary.

Corollary 3. (A direct result following [17]) For the state-dependent IC with states noncausally known at both transmitters, if $|a(1+b^2P_1)| + |b(1+a^2P_2)| \le 1$, then the sum capacity is given by

$$C_{sum} = \frac{1}{2} \log \left(1 + \frac{P_1}{a^2 P_2 + 1} \right) + \frac{1}{2} \log \left(1 + \frac{P_2}{b^2 P_1 + 1} \right).$$

For the state-dependent Z-IC with states noncausally known at both transmitters, if $a^2 \le 1$, then the sum capacity is given by

$$C_{sum} = \frac{1}{2} \log \left(1 + \frac{P_1}{a^2 P_2 + 1} \right) + \frac{1}{2} \log \left(1 + P_2 \right).$$

It can be seen that the sum capacity achieving scheme does not depend on the correlation of the states, and hence, in the weak interference regime, the sum capacity is not affected by the correlation of the states.

VI. STATE-DEPENDENT IC WITH PARTIAL STATE INFORMATION

In this section, we study the state-dependent IC with each state only known at the corresponding transmitter. Differently from the model studied in Sections III, IV and V, where both state interferences are known to both transmitters, the partial state cognition is challenging because the transmitters cannot cooperate with each other via their states due to the lack of the other transmitter's state information. We investigate the very strong/strong and weak interference regimes for both the state-dependent regular IC and Z-IC.

A. State-Dependent Regular IC

For the state-dependent regular IC, without the full knowledge of the channel states, we develop a new achievable scheme presented below, in which each transmitter encodes its message based only on its own knowledge about the channel state and the interference.

Proposition 5. For the state-dependent IC with state non-causally known at corresponding transmitter, the achievable region consists of rate pairs (R_1, R_2) satisfying:

$$R_1 \leq \min\{I(U; VY_1), I(U; Y_2)\} - I(S_1; U),$$
 (35a)

$$R_2 \leq \min\{I(V; UY_2), I(V; Y_1)\} - I(S_2; V)$$
 (35b)

for some distribution $P_{S_1S_2}P_{U|S_1}P_{X_1|US_1}P_{V|S_2}P_{X_2|VS_2}$ $P_{Y_1Y_2|S_1S_2X_1X_2}$, where U and V are auxiliary random variables.

We next use the Gaussian IC defined in (2a) and (2b) to illustrate the above achievable region numerically. Specifically, since the state is not know to both transmitters, the channel is no longer equivalent to (6a) and (6b). Hence, the input variables X_1 and X_2 can no longer be generated based on both S_1' and S_2 . Thus, dirty paper coding is no longer optimal, and the achievable region relies on the result of a new optimization problem.

Proposition 6. For the state-dependent Gaussian IC with each state noncausally known at the corresponding transmitter, an inner bound on the capacity region consists of rate pairs (R_1, R_2) satisfying (36a) and (36b), as shown at the top of the next page, where $-1 \le \rho \le 1$.

Proof: The proof follows from Proposition 5 by setting $U = X_1 + \alpha S_1$, $V = X_2 + \beta S_2$, $X_1 \sim \mathcal{N}(0, P_1)$, $X_2 \sim \mathcal{N}(0, P_2)$, $S_1 \sim \mathcal{N}(0, Q_1)$, $S_2 \sim \mathcal{N}(0, Q_2)$ and $Cov(S_1, S_2) = \rho \sqrt{Q_1 Q_2}$.

When the state is known only to corresponding transmitter, the encoders cannot encoding with fully utilize of the state correlation and cannot optimize the decoding capability for both receivers. Therefore, the coding scheme is based on dirty paper coding for the whole state, rather than dealing with the correlated state and independent state respectively. In this case, the interference from the other transmitter has limited assistance in canceling the state, and even under very strong or strong interference regime, the interference cannot be fully decoded and cancelled. Based on the achievable region in Proposition 6, the interference and state cannot be fully cancelled, especially when the correlation between states is very small. This also suggests that the outer bound is related with the correlation, and is smaller than the very strong/strong interference channel without state.

For the weak interference regime, as explained in Section V, each transmitter can still encode based on dirty paper coding with respect to its corresponding state, and treat the interference as noise, so that the sum capacity of the channel without the state can be achieved, which is thus the sum capacity of the state-dependent channel as stated in Corollary 3. In such a case, though the transmitters have only partial information about the state, they can still fully cancel the state interference at their corresponding receiver.

B. State-Dependent Z-IC

In this subsection, we further investigate the Z-IC with each state known only to its corresponding transmitter. For the state-dependent Z-IC, the encoding scheme is the same as that for the regular IC. In the decoding part, since only one receiver is corrupted by the interference, the interference-free receiver decodes only its own auxiliary random variable. Hence, the achievable region is given as follows.

Proposition 7. For the state-dependent Z-IC with each state noncausally known at its corresponding transmitter,

the achievable region consists of rate pairs (R_1, R_2) satisfying:

$$R_1 \leqslant I(U; VY_1) - I(S_1; U),$$
 (37a)

$$R_2 \leq \min\{I(V; UY_2), I(V; Y_1)\} - I(S_2; V)$$
 (37b)

for some distribution $P_{S_1S_2}P_{U|S_1}P_{X_1|US_1}P_{V|S_2}P_{X_2|VS_2}$ $P_{Y_1Y_2|S_1S_2X_1X_2}$, where U and V are auxiliary random variables.

We use the same channel model as in (2a) and (2b) by setting b=0, i.e., receiver 2 is interference-free. And we obtain the achievable region as follows.

Proposition 8. For the state-dependent Gaussian IC with only corresponding state noncausally known at each transmitter, an inner bound on the capacity region consists of rate pairs (R_1, R_2) satisfying (38a) and (38b), as shown at the top of the next page, where $-1 \le \rho \le 1$.

Proof: The proof follows from Proposition 7 with the same distribution described in the proof of Proposition 6. \Box

VII. CONCLUSION

In this paper, we studied the state-dependent Gaussian IC and Z-IC with receivers being corrupted by two correlated states which are noncausally known at transmitters. We explored two cases respectively with state noncausally known at both transmitters and with each state noncausally know at only its corresponding transmitter. For the first case, we characterized the conditions on the channel parameters under which state-dependent IC and Z-IC achieve the capacity region or the sum capacity of the corresponding channel without state. Our result suggests that more correlated states tend to make it easier to fully cancel the states. Our comparison between the IC and the Z-IC suggests that the IC benefits more if the correlation between the states increases. The second case turns out to be substantially challenging to analyze, and we derived the achievable regions for the regular IC and Z-IC. Characterization of the capacity region either partially or fully is still an open problem. We anticipates that the state cancellation schemes we develop here can be useful for studying other state-dependent models. We also expect that simple schemes such as the layered approach designed for the interference channel together with dirty-paper coding can perform very well, when these schemes are tailored for practical communication schemes.

APPENDIX A PROOF OF PROPOSITION 1

We use random codes and fix the following joint distribution: $P_{S_1S_2UVX_1X_2Y_1Y_2} = P_{S_1S_2}P_{U|S_1S_2}P_{X_1|US_1S_2}P_{V|S_1S_2}P_{X_2|VS_1S_2}P_{Y_1Y_2|S_1S_2X_1X_2}$.

- 1) Codebook Generation:
 - Generate $2^{n(R_1+\tilde{R}_1)}$ codewords $U^n(w_1,l_1)$ with i.i.d. components based on P_U . Index these codewords by $w_1=1,\ldots,2^{nR_1},l_1=1,\ldots,2^{n\tilde{R}_1}$.
 - Generate $2^{n(R_2+\tilde{R}_2)}$ codewords $V^n(w_2,l_2)$ with i.i.d. components based on P_V . Index these codewords by $w_2=1,\ldots,2^{nR_2},l_2=1,\ldots,2^{n\tilde{R}_2}$.

$$\begin{split} R_{1}(\alpha,\beta) &\leq \min \Big\{ \frac{1}{2} \log \frac{P_{1}(P_{2} + b^{2}P_{1} + Q_{2} + 1)}{(P_{2} + 1)(P_{1} + \alpha^{2}Q_{1}) + (1 - \rho)^{2}\alpha^{2}Q_{1}(b^{2}P_{1} + Q_{2}) + P_{1}(\alpha\rho b\sqrt{Q_{1}} - \sqrt{Q_{2}})^{2}}, \\ &\frac{1}{2} \log \frac{P_{1}((P_{1} + 1)(P_{2} + \beta^{2}Q_{2}) + (1 - \rho^{2})\beta^{2}Q_{2}(Q_{1} + a^{2}P_{2}) + P_{2}(\sqrt{Q_{1}} - a\beta\rho\sqrt{Q_{2}})^{2})}{P_{1}Q_{1}(1 - \alpha)^{2}(P_{2} + (1 - \rho^{2})\beta^{2}Q_{2}) + P_{2}(P_{1} + \alpha^{2}Q_{1}) + (a^{2}P_{2} + 1)\beta^{2}Q_{2}(P_{1} + (1 - \rho^{2})\alpha^{2}Q_{1}) + 2a\beta\rho P_{1}P_{2}\sqrt{Q_{1}Q_{2}}(\alpha - 1)} \Big\} \end{split}$$

$$(36a)$$

$$\begin{split} R_{2}(\alpha,\beta) &\leq \min \Big\{ \frac{1}{2} \log \frac{P_{2}(P_{1} + a^{2}P_{2} + Q_{1} + 1)}{(P_{1} + 1)(P_{2} + \beta^{2}Q_{2}) + (1 - \rho)^{2}\beta^{2}Q_{2}(a^{2}P_{2} + Q_{1}) + P_{2}(\beta\rho a\sqrt{Q_{2}} - \sqrt{Q_{1}})^{2}}, \\ &\frac{1}{2} \log \frac{P_{2}((P_{2} + 1)(P_{1} + \alpha^{2}Q_{1}) + (1 - \rho^{2})\alpha^{2}Q_{1}(Q_{2} + b^{2}P_{1}) + P_{1}(\sqrt{Q_{2}} - b\alpha\rho\sqrt{Q_{1}})^{2})}{P_{2}Q_{2}(1 - \beta)^{2}(P_{1} + (1 - \rho^{2})\alpha^{2}Q_{1}) + P_{1}(P_{2} + \beta^{2}Q_{2}) + (b^{2}P_{1} + 1)\alpha^{2}Q_{1}(P_{2} + (1 - \rho^{2})\beta^{2}Q_{2}) + 2b\alpha\rho P_{1}P_{2}\sqrt{Q_{1}Q_{2}}(\beta - 1)} \Big\} \end{split}$$
(36b)

$$R_1(\alpha, \beta)$$

$$\leq \frac{1}{2} \log \frac{P_1((P_1+1)(P_2+\beta^2Q_2)+(1-\rho^2)\beta^2Q_2(Q_1+a^2P_2)+P_2(\sqrt{Q_1}-a\beta\rho\sqrt{Q_2})^2)}{P_1Q_1(1-\alpha)^2(P_2+(1-\rho^2)\beta^2Q_2)+P_2(P_1+\alpha^2Q_1)+(a^2P_2+1)\beta^2Q_2(P_1+(1-\rho^2)\alpha^2Q_1)+2a\beta\rho P_1P_2\sqrt{Q_1Q_2}(\alpha-1)}$$
(38a)

$$R_{2}(\alpha,\beta) \leq \min \left\{ \frac{1}{2} \log \frac{P_{2}(P_{1} + a^{2}P_{2} + Q_{1} + 1)}{(P_{1} + 1)(P_{2} + \beta^{2}Q_{2}) + (1 - \rho)^{2}\beta^{2}Q_{2}(a^{2}P_{2} + Q_{1}) + P_{2}(\beta\rho a\sqrt{Q_{2}} - \sqrt{Q_{1}})^{2}}, \frac{1}{2} \log \frac{P_{2}((P_{2} + 1)(P_{1} + \alpha^{2}Q_{1}) + (1 - \rho^{2})\alpha^{2}Q_{1}(Q_{2} + b^{2}P_{1}) + P_{1}(\sqrt{Q_{2}} - b\alpha\rho\sqrt{Q_{1}})^{2})}{P_{2}Q_{2}(1 - \beta)^{2}(P_{1} + (1 - \rho^{2})\alpha^{2}Q_{1}) + P_{1}(P_{2} + \beta^{2}Q_{2}) + (b^{2}P_{1} + 1)\alpha^{2}Q_{1}(P_{2} + (1 - \rho^{2})\beta^{2}Q_{2}) + 2b\alpha\rho P_{1}P_{2}\sqrt{Q_{1}Q_{2}}(\beta - 1)} \right\}$$
(38b)

2) Encoding:

• Transmitter 1: Given (s_1^n, s_2^n) and w_1 , choose a $u^n(w_1, \tilde{l}_1)$ such that

$$(u^n(w_1, \tilde{l}_1), s_1^n, s_2^n) \in T_{\epsilon}^n(P_{S_1S_2U}).$$

Otherwise, set $\tilde{l}_1 = 1$. It can be shown that for large n, such u^n exists with high probability if

$$\tilde{R}_1 > I(U; S_1 S_2).$$
 (39)

Then generate x_1^n with i.i.d. component based on $P_{X_1|US_1S_2}$ for transmission.

• Transmitter 2: Given (s_1^n, s_2^n) and w_2 , choose a $v^n(w_2, \tilde{l}_2)$ such that

$$(v^n(w_2, \tilde{l}_2), s_1^n, s_2^n) \in T_{\epsilon}^n(P_{S_1S_2V}).$$

Otherwise, set $\tilde{l}_2 = 1$. It can be shown that for large n, such v^n exists with high probability if

$$\tilde{R}_2 > I(V; S_1 S_2).$$
 (40)

Then generate x_2^n with i.i.d. components based on $P_{X_2|VS_1S_2}$ for transmission.

3) Decoding:

• Decoder 1: Given y_1^n , find (\hat{w}_2, \hat{l}_2) such that

$$(v^n(w_2, \hat{l}_2), y_1^n) \in T_{\epsilon}^n(P_{VY_1}).$$

If no or more than one such pair (\hat{w}_2, \hat{l}_2) can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_2 + \tilde{R}_2 \le I(V; Y_1).$$
 (41)

After decoding v^n , find a unique pair (\hat{w}_1, \hat{l}_1) such that

$$(u^n(\hat{w}_1,\hat{l}_1),v^n(w_2,\hat{l}_2),y_1^n)\in T_{\epsilon}^n(P_{VUY_1}).$$

If no or more than one such pair (\hat{w}_2, \hat{l}_2) can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_1 + \tilde{R}_1 \leqslant I(U; VY_1). \tag{42}$$

• Decoder 2: Given y_2^n , find (\hat{w}_1, \hat{l}_1) such that

$$(u^n(w_1,\hat{l}_1),y_2^n)\in T_c^n(P_{UY_2}).$$

If no or more than one such pair (\hat{w}_1, \hat{l}_1) can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_1 + \tilde{R}_1 \le I(U; Y_2).$$
 (43)

After decoding u^n , find a unique pair (\hat{w}_2, \hat{l}_2) such that

$$(v^n(\hat{w}_2,\hat{l}_2), u^n(w_1,\hat{l}_1), y_2^n) \in T_{\epsilon}^n(P_{VUY_2}).$$

If no or more than one such pair can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_2 + \tilde{R}_2 \leqslant I(V; UY_2). \tag{44}$$

Proposition 1 is thus proved by combining (39)-(44).

APPENDIX B PROOF OF PROPOSITION 2

We use random codes and fix the following joint distribution: $P_{S_1S_2UVX_1X_2Y_1Y_2} = P_{S_1S_2}P_{U|S_1S_2}P_{X_1|US_1S_2}P_{V|S_1S_2}P_{X_2|VS_1S_2}P_{Y_1Y_2|S_1S_2X_1X_2}$.

1) Codebook Generation:

- Generate $2^{n(R_1+\tilde{R}_1)}$ codewords $U^n(w_1,l_1)$ with i.i.d. components based on P_U . Index these codewords by $w_1=1,\ldots,2^{nR_1},l_1=1,\ldots,2^{n\tilde{R}_1}$.
- Generate $2^{n(R_2+\tilde{R}_2)}$ codewords $V^n(w_2, l_2)$ with i.i.d. components based on P_V . Index these codewords by $w_2 = 1, \dots, 2^{nR_2}, l_2 = 1, \dots, 2^{n\tilde{R}_2}$.

2) Encoding:

• Transmitter 1: Given (s_1^n, s_2^n) and w_1 , choose a $u^n(w_1, \tilde{l}_1)$ such that

$$(u^n(w_1, \tilde{l}_1), s_1^n, s_2^n) \in T_{\epsilon}^n(P_{S_1S_2U}).$$

Otherwise, set $\tilde{l}_1 = 1$. It can be shown that for large n, such u^n exists with high probability if

$$\tilde{R}_1 > I(U; S_1 S_2).$$
 (45)

Then generate x_1^n with i.i.d. component based on $P_{X_1|US_1S_2}$ for transmission.

• Transmitter 2: Given (s_1^n, s_2^n) and w_2 , choose a $v^n(w_2, \tilde{l}_2)$ such that

$$(v^n(w_2, \tilde{l}_2), s_1^n, s_2^n) \in T_{\epsilon}^n(P_{S_1S_2V}).$$

Otherwise, set $\tilde{l}_2 = 1$. It can be shown that for large n, such v^n exists with high probability if

$$\tilde{R}_2 > I(V; S_1 S_2).$$
 (46)

Then generate x_2^n with i.i.d. component based on $P_{X_2|VS_1S_2}$ for transmission.

3) Decoding:

• Decoder 1: Given y_1^n , find (\hat{w}_2, \hat{l}_2) such that

$$(v^n(\hat{w}_2,\hat{l}_2), y_1^n) \in T_c^n(P_{VY_1}).$$

If no or more than one such pair (\hat{w}_2, \hat{l}_2) can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_2 + \tilde{R}_2 \leqslant I(V; Y_1).$$
 (47)

After decoding v^n , find a unique pair (\hat{w}_1, \hat{l}_1) such that

$$(u^n(\hat{w}_1, \hat{l}_1), v^n(w_2, \hat{l}_2), y_1^n) \in T_{\epsilon}^n(P_{VUY_1}).$$

If no or more than one such pair can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_1 + \tilde{R}_1 \leqslant I(U; VY_1).$$
 (48)

• Decoder 2: Given y_2^n , find (\hat{w}_2, \hat{l}_2) such that

$$(v^n(w_2,\hat{l}_1), y_2^n) \in T_{\epsilon}^n(P_{UY_2}).$$

If no or more than one such pair (\hat{w}_2, \hat{l}_2) can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_2 + \tilde{R}_2 \leqslant I(V; Y_2).$$
 (49)

Proposition 2 is thus proved by combining (45)-(49).

APPENDIX C PROOF OF PROPOSITION 3

We use random codes and fix the following joint distribution: $P_{S_1S_2U_1U_2VX_1X_2Y_1Y_2} = P_{S_1S_2}P_{V|S_1}P_{X_2|VS_1}P_{U_1|S_1}P_{U_2|S_1U_1}P_{X_1|U_1U_2S_1}P_{Y_1|S_1X_1X_2}P_{Y_2|S_2X_2}$.

1) Codebook Generation:

- Generate $2^{n(R_{11}+\tilde{R}_{11})}$ codewords $U_1^n(w_{11},l_{11})$ with i.i.d. components based on P_{U_1} . Index these codewords by $w_{11}=1,\ldots,2^{nR_{11}},l_{11}=1,\ldots,2^{n\tilde{R}_{11}}$.
- For each $u_1^n(w_{11}, l_{11})$, generate $2^{n(R_{12} + \tilde{R}_{12})}$ codewords $U_2^n(w_{11}, l_{11}, w_{12}, l_{12})$ with i.i.d. components based on $P_{U_2|U_1}$. Index these codewords by $w_{12} = 1, \ldots, 2^{nR_{12}}, l_{12} = 1, \ldots, 2^{n\tilde{R}_{12}}$.
- Generate $2^{n(R_2+\tilde{R}_2)}$ codewords $V^n(w_2,l_2)$ with i.i.d. components based on P_V . Index these codewords by $w_2=1,\ldots,2^{nR_2}, v=1,\ldots,2^{n\tilde{R}_2}$.

2) Encoding:

• Transmitter 1: Given s_1^n and w_{11} , choose a $u_1^n(w_{11}, \tilde{l}_{11})$ such that

$$(u^n(w_{11}, \tilde{l}_{11}), s_1^n) \in T_{\epsilon}^n(P_{S_1U_{11}}).$$

Otherwise, set $\tilde{l}_{11} = 1$. It can be shown that for large n, such u_1^n exists with high probability if

$$\tilde{R}_{11} > I(U_1; S_1).$$
 (50)

Given w_{11} , \tilde{l}_{11} , w_{12} and s_1^n , choose a $u_2^n(w_{11}, \tilde{l}_{11}, w_{12}, \tilde{l}_{12})$ such that

$$(u_1^n(w_{11}, \tilde{l}_{11}), u_2^n(w_{11}, \tilde{l}_{11}, w_{12}, \tilde{l}_{12}), s_1^n) \\ \in T_{\epsilon}^n(P_{S_1U_1U_2}).$$

Otherwise, set $\tilde{l}_{12} = 1$. It can be shown that for large n, such u_2^n exists with high probability if

$$\tilde{R}_{12} > I(U_2; S_1 | U_1).$$
 (51)

Given $u_1^n(w_{11}, \tilde{l}_{11})$, $u_2^n(w_{11}, \tilde{l}_{11}, w_{12}, \tilde{l}_{12})$ and s_1^n , generate x_1^n with i.i.d. components based on $P_{X_1|S_1U_1U_2}$.

• Transmitter 2: Given s_1^n and w_2 , choose a $v^n(w_2, \tilde{l}_2)$ such that

$$(v^n(w_2, \tilde{l}_2), s_1^n) \in T_{\epsilon}^n(P_{S_1V}).$$

Otherwise, set $\tilde{l}_2 = 1$. It can be shown that for large n, such v^n exists with high probability if

$$\tilde{R}_2 > I(V; S_1). \tag{52}$$

Then generate x_2^n with i.i.d. component based on $P_{X_2|VS_1}$ for transmission.

3) Decoding:

• Decoder 1: Given y_1^n , find $(\hat{w}_{11}, \hat{l}_{11})$ such that

$$(u_1^n(\hat{w}_{11},\hat{l}_{11}),y_1^n)\in T_{\epsilon}^n(P_{U_1Y_1}).$$

If no or more than one such a pair $(\hat{w}_{11}, \hat{l}_{11})$ can be found, declare an error. It is easy to show that

for sufficiently large n, we can correctly find such a pair with high probability if

$$R_{11} + \tilde{R}_{11} \leqslant I(U_1; Y_1).$$
 (53)

After decoding u_1^n , find a unique pair (\hat{w}_2, \hat{l}_2) such that

$$(u_1^n(\hat{w}_{11},\hat{l}_{11}),v^n(w_2,\hat{l}_2),y_1^n)\in T_{\epsilon}^n(P_{VU_1Y_1}).$$

If no or more than one such pair can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_2 + \tilde{R}_2 \leqslant I(V; Y_1 | U_1).$$
 (54)

After successively decoding v^n , find a unique tuple $(w_{11}, \tilde{l}_{11}, w_{12}, \tilde{l}_{12})$ such that $((u_1^n(\hat{w}_{11}, \hat{l}_{11}), v^n(w_2, \hat{l}_2), u_2^n(w_{11}, \tilde{l}_{11}, w_{12}, \tilde{l}_{12}), y_1^n) \in T_{\epsilon}^n(P_{VU_1U_2Y_1}))$. If no or more than one such pair can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_{12} + \tilde{R}_{12} \le I(U_2; VY_1|U_1).$$
 (55)

• Decoder 2: Given y_2^n , find $(\hat{w}_{11}, \hat{l}_{11})$ such that

$$(u_1^n(\hat{w}_{11},\hat{l}_{11}),y_2^n)\in T_{\epsilon}^n(P_{U_1Y_1}).$$

If no or more than one such pair $(\hat{w}_{11}, \hat{l}_{11})$ can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_{11} + \tilde{R}_{11} \leqslant I(U_1; Y_2).$$
 (56)

After decoding u_1^n , find a unique pair (\hat{w}_2, \hat{l}_2) such that

$$(u_1^n(\hat{w}_{11},\hat{l}_{11}),v^n(w_2,\hat{l}_2),y_2^n)\in T_{\epsilon}^n(P_{VU_1Y_2}).$$

If no or more than one such pair can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_2 + \tilde{R}_2 \le I(V; Y_2 | U_1).$$
 (57)

After successively decoding v^n , find a unique tuple $(w_{11}, \tilde{l}_{11}, w_{12}, \tilde{l}_{12})$ such that $((u_1^n(\hat{w}_{11}, \hat{l}_{11}), v^n(w_2, \hat{l}_2), u_2^n(w_{11}, \tilde{l}_{11}, w_{12}, \tilde{l}_{12}), y_1^n) \in T_{\epsilon}^n(P_{VU_1U_2Y_2}))$. If no or more than one such pair can be found, declare an error. It is easy to show that for sufficiently large n, we can correctly find such a pair with high probability if

$$R_{12} + \tilde{R}_{12} \leqslant I(U_2; VY_2|U_1).$$
 (58)

The corresponding achievable region is thus characterized by

$$R_{11} \leq \min\{I(U_1; Y_1), I(U_1; Y_2)\} - I(U_1; S_1)$$
 (59)

$$R_{12} \leq \min\{I(U_2; VY_1|U_1), I(U_2; VY_2|U_1)\}$$

$$-I(U_2; S_1|U_1) (60)$$

$$R_2 \leq \min\{I(V; Y_1|U_1), I(V; Y_2|U_1)\} - I(V; S_1).$$

Proposition 3 is completed by combining (59)-(61) and $R_1 = R_{11} + R_{12}$.

APPENDIX D PROOF OF PROPOSITION 5

We use random codes and fix the following joint distribution: $P_{S_1S_2UVX_1X_2Y_1Y_2} = P_{S_1S_2}P_{U|S_1}P_{X_1|US_1}P_{V|S_2}P_{Y_1Y_2|S_1S_2X_1X_2}$.

- 1) Codebook Generation:
 - Generate 2^{n(R₁+K̄₁)} codewords Uⁿ(w₁, l₁) with i.i.d. components based on P_U. Index these codewords by w₁ = 1,..., 2^{nR₁}, l₁ = 1,..., 2^{nK̄₁}.
 Generate 2^{n(R₂+K̄₂)} codewords Vⁿ(w₂, l₂) with i.i.d.
 - Generate $2^{n(R_2+R_2)}$ codewords $V^n(w_2, l_2)$ with i.i.d. components based on P_V . Index these codewords by $w_2 = 1, \dots, 2^{nR_2}, l_2 = 1, \dots, 2^{n\tilde{R}_2}$.
- 2) Encoding:
 - Transmitter 1: Given (s_1^n) and w_1 , choose a $u^n(w_1, \tilde{l}_1)$ such that

$$(u^n(w_1, \tilde{l}_1), s_1^n) \in T_{\epsilon}^n(P_{US_1}).$$

Otherwise, set $\tilde{l}_1 = 1$. It can be shown that for large n, such u^n exists with high probability if

$$\tilde{R}_1 > I(U; S_1). \tag{62}$$

Then generate x_1^n with i.i.d. component based on $P_{X_1|US_1}$ for transmission.

• Transmitter 2: Given (s_2^n) and w_2 , choose a $v^n(w_2, \tilde{l}_2)$ such that

$$(v^n(w_2, \tilde{l}_2), s_2^n) \in T_{\epsilon}^n(P_{VS_2}).$$

Otherwise, set $\tilde{l}_2 = 1$. It can be shown that for large n, such v^n exists with high probability if

$$\tilde{R}_2 > I(V; S_2). \tag{63}$$

Then generate x_2^n with i.i.d. components based on $P_{X_2|VS_2}$ for transmission.

3) Decoding: The decoding procedure is the same as in Appendix A and is hence omitted.

Proposition 5 is thus proved by combining (62), (63) and (41)–(44).

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