

On Capacities of the Two-User Union Channel With Complete Feedback

Zilin Jiang^{ID}, Nikita Polyanskii^{ID}, and Ilya Vorobyev^{ID}

Abstract—The exact values of the optimal symmetric rate point in the Cover–Leung capacity region of the two-user union channel with complete feedback were determined by Willems when the size of the input alphabet is 2, and by Vinck *et al.* when the size is at least 6. We complete this line of research when the size of the input alphabet is 3, 4, or 5. The proof hinges on the technical lemma that concerns the maximal joint entropy of two independent random variables in terms of their probability of equality. For the zero-error capacity region, using superposition coding, we provide a practical near-optimal communication scheme which improves all the previous explicit constructions.

Index Terms—Union channel, feedback, channel capacity, zero-error capacity, entropy function.

I. INTRODUCTION

THE two-user union channel, first introduced in [1] and rediscovered in [2], is a discrete memoryless multiple-access channel¹: the channel takes symbols x_1, x_2 from the input alphabet $\mathcal{X} := [q] = \{1, 2, \dots, q\}$ given by two senders, and outputs the union $y = \{x_1, x_2\}$ from the output alphabet $\mathcal{Y} := \{y \subseteq [q] : |y| \in \{1, 2\}\}$. For the special case $q = 2$, the union channel coincides with the *two-user binary adder channel*.

Since a received $y \in \mathcal{Y}$ cannot be unambiguously decoded, the central problem in two-user communication theory is to coordinate the two senders to send simultaneously as much information as possible to a single receiver through n uses of the union channel.

Manuscript received April 23, 2018; revised September 22, 2018; accepted November 27, 2018. Z. Jiang was supported by the Israel Science Foundation (ISF) under Grant 1162/15 and Grant 936/16. N. Polyanskii was supported in part by ISF under Grant 1162/15 and Grant 326/17 and in part by the Russian Foundation for Basic Research (RFBR) under Grant 16-01-00440 A, Grant 18-07-01427 A, and Grant 18-31-00310 MOL_A. I. Vorobyev was supported by RFBR under Grant 16-01-00440 A, Grant 18-07-01427 A, and Grant 18-31-00361 MOL_A.

Z. Jiang was with the Technion–Israel Institute of Technology, Haifa 3200003, Israel. He is now with the Department of Mathematics, Massachusetts Institute of Technology, Cambridge, MA 02139 USA (e-mail: zilinj@mit.edu).

N. Polyanskii is with CDISE, Skolkovo Institute of Science and Technology, 121205 Moscow, Russia, and also with the Department of Mathematics, Technion–Israel Institute of Technology, Haifa 3200003, Israel (e-mail: nikita.polyansky@gmail.com).

I. Vorobyev is with CDISE, Skolkovo Institute of Science and Technology, 121205 Moscow, Russia, and also with the Moscow Institute of Physics and Technology, 141701 Dolgoprudny, Russia (e-mail: vorobyev.i.v@yandex.ru).

Communicated by M. Lentmaier, Associate Editor for Coding Theory.

Digital Object Identifier 10.1109/TIT.2018.2889250

¹The terminology from information theory used throughout the article is standard, and can be found in [3].

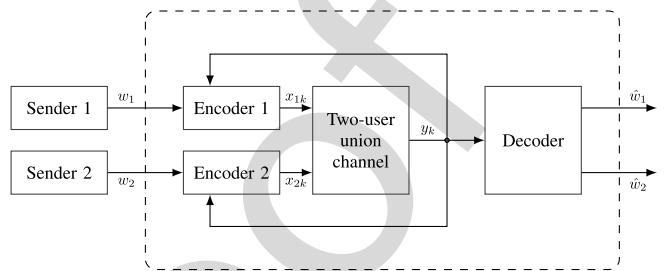


Fig. 1. Two-user union channel with complete feedback.

Let the message sets specified for the senders be of size M_1 and M_2 , and let $w_1 \in [M_1]$, $w_2 \in [M_2]$ be two messages chosen by the two senders beforehand. During the k th use of the channel, two functions e_{1k} and e_{2k} respectively encode w_1 and w_2 to two codewords $x_{1k} \in [q]$ and $x_{2k} \in [q]$. The union channel then takes x_{1k}, x_{2k} and outputs $y_k := \{x_{1k}, x_{2k}\} \in \mathcal{Y}$. The sequence of outputs $(y_k)_{k=1}^n$ is decoded by the receiver to the estimate (\hat{w}_1, \hat{w}_2) of (w_1, w_2) .

In such a communication scheme for the transmission of information from two senders to one receiver, the encoders at any moment might know some information about the signals received by the decoder prior to this moment. When the encoders know nothing but the message from their corresponding senders, we say that the channel is the *two-user union channel without feedback*. However, when the encoders also know all the previous outputs of the channel, namely every e_{1k} and e_{2k} depend not only on w_1 and w_2 respectively but also on $(y_i)_{i=1}^{k-1}$, we say that the channel is the *two-user union channel with complete feedback*. See [4, Sec. 4] for a broader view on coding for the multiple-access channel.

In this work, we mainly focus on capacities of the two-user union channel with complete feedback. An $(M_1, M_2, n, \varepsilon)$ code for the two-user union channel with complete feedback consists of a collection encoding functions and a decoding function such that the *probability of error*, defined by $\Pr((\hat{w}_1, \hat{w}_2) \neq (w_1, w_2))$ when (w_1, w_2) is drawn uniformly from $[M_1] \times [M_2]$, is at most ε . In particular, an $(M_1, M_2, n, 0)$ code could recover the messages without errors. The channel capacity region \mathcal{E}_f for the two-user union channel with complete feedback captures the rates at which the information can be transmitted over the channel for both users with arbitrarily small probability of error, whereas the zero-error capacity region \mathcal{O}_f represents the rates without error:

$$\mathcal{E}_f := \text{closure of } \{(R_1, R_2) : \exists \text{ a sequence of } ([q^{nR_1}], [q^{nR_2}], n, \varepsilon_n) \text{ codes s.t. } \varepsilon_n \rightarrow 0\},$$

63 $\mathcal{O}_f := \text{closure of } \{(R_1, R_2) : \exists \text{ a sequence of}$
 64 $(\lceil q^{nR_1} \rceil, \lceil q^{nR_2} \rceil, n, \varepsilon_n) \text{ codes s.t. } \varepsilon_n = 0 \text{ eventually}\}$.

65 In the absence of feedback, the channel capacity region \mathcal{E} and
 66 the zero-error capacity region \mathcal{O} are similarly defined for the
 67 two-user union channel.

68 For each of the above capacity regions, say \mathcal{C} , research has
 69 been devoted to the *average capacity*

$$70 R(\mathcal{C}) := \sup \left\{ \frac{1}{2}(R_1 + R_2) : (R_1, R_2) \in \mathcal{C} \right\},$$

71 which can be understood as the maximal rate per user at which
 72 the information can be transmitted. Because \mathcal{C} is convex and
 73 symmetric with respect to the line $R_1 = R_2$, the average
 74 capacity $R(\mathcal{C})$ can also be defined as $\sup \{R : (R, R) \in \mathcal{C}\}$.
 75 The point $(R(\mathcal{C}), R(\mathcal{C}))$ is known as the equal-rate point or
 76 the symmetric rate point in the existing literature.

77 The channel capacity region for a discrete memory-
 78 less multiple-access channel without feedback has been
 79 fully characterized by Alswede [5] and Liao [6]. For the
 80 two-user union channel without feedback, the average chan-
 81 nel capacity $R(\mathcal{E}) = 1 - \frac{q-1}{2q \log_2 q}$ has been determined
 82 by Chang and Wolf [1]. Much less is known for the aver-
 83 age zero-error capacity $R(\mathcal{O})$ of the two-user union chan-
 84 nel without feedback. For $q = 2$, there is no better
 85 upper bound other than the trivial $R(\mathcal{O}) \leq R(\mathcal{E}) =$
 86 0.75, while the current record lower bound is $R(\mathcal{O}) \geq$
 87 $\frac{1}{12} \log_2 240 = 0.65891$ due to Mattas and Östergård [7,
 88 Sec. III] obtained by computer searches. For $q \geq 3$, several
 89 constructions provided by Chang and Wolf [1, Sec. III] imply
 90 that $R(\mathcal{O}) \geq \frac{1}{4}(1 + \log_q(q^2 - q + 1))$ for all $q \geq 2$, $R(\mathcal{O}) \geq$
 91 $\log_q(\frac{1}{2}(q+1))$ for odd q and $R(\mathcal{O}) \geq \log_q(\frac{1}{2}\sqrt{q(q+2)})$ for
 92 even q . The variation where the senders are required to use
 93 the same encoding functions was studied in various context.
 94 We refer the readers to [8] for the best code construction when
 95 the size of the input alphabet is 2, to [9] for the connection
 96 with the binary B_2 -sequences, and to [10] for large input
 97 alphabet. The generalization, in which more than 2 users
 98 have access to the channel, was recently investigated in [11]
 99 and [12].

100 Gaarder and Wolf [13] demonstrated that feedback may
 101 increase the channel capacity region. They used the two-user
 102 binary adder channel as an example and developed a simple
 103 two-stage coding strategy. Using the concept of superposi-
 104 tion coding Cover and Leung [14] characterized a subset of
 105 the channel capacity region \mathcal{E}_f for the discrete memoryless
 106 multiple-access channels with complete feedback. This subset
 107 was later shown to be exactly \mathcal{E}_f by Willems [15] for the class
 108 of the channels where one of the inputs is determined by the
 109 other input and the output. Their results are paraphrased as
 110 the following theorem in the special case that the channel is
 111 the two-user union channel.

112 **Theorem 1 (Theorem 1 of Cover and Leung [14] and The-
 113 orem of Willems [15]):** The channel capacity region \mathcal{E}_f of the
 114 two-user union channel using input alphabet $[q]$ with complete
 115 feedback is the convex hull of all (R_1, R_2) satisfying

$$116 \quad 0 \leq R_1 \leq H(X_1 | U), \quad 0 \leq R_2 \leq H(X_2 | U), \\ 117 \quad R_1 + R_2 \leq H(\{X_1, X_2\})$$

where U is a discrete random variable,² X_1, X_2 are two
 $[q]$ -valued random variables that are conditionally independent
 given U , and the entropy function H uses the base- q
 logarithm.

122 **Remark 1:** The entropy $H(\{X_1, X_2\})$ in Theorem 1 is the
 123 entropy of the random variable $Y := \{X_1, X_2\}$, not to be
 124 confused with the joint entropy $H(X_1, X_2)$.

125 For $q = 2$, Willems [17] later showed that $R(\mathcal{E}_f) =$
 126 0.79113; for $q \geq 6$, Vinck *et al.* [2] asserted that $R(\mathcal{E}_f) =$
 127 $\frac{1}{2} \log_q \binom{q+1}{2}$. In Section II we complete this line of research on
 128 $R(\mathcal{E}_f)$ for all $q \geq 2$. The proof hinges on the following lemma
 129 about the maximum of the joint entropy of two independent
 130 discrete random variables in terms of their probability of
 131 equality.

132 **Lemma 2:** Given $q \geq 2$, for every $\theta \in [0, 1]$ let $F(\theta)$ be
 133 the maximum of the joint entropy $H(X_1, X_2)$ among all pairs
 134 of independent $[q]$ -valued random variables X_1, X_2 such that
 135 $\Pr(X_1 = X_2) = \theta$. The function $F: [0, 1] \rightarrow \mathbb{R}$ is continuous
 136 and it is increasing on $[0, 1/q]$ and decreasing on $[1/q, 1]$.
 137 Moreover,

$$138 \quad F(\theta) = 2(-\alpha \log \alpha - (1-\alpha) \log(1-\alpha) \\ 139 \quad + (1-\alpha) \log(q-1)), \text{ for } \theta \in [1/q, 1]$$

140 where the bijection $\alpha: [1/q, 1] \rightarrow [1/q, 1]$ is defined by

$$141 \quad \alpha = \alpha(\theta) := \frac{1}{q} + \sqrt{\left(1 - \frac{1}{q}\right)\left(\theta - \frac{1}{q}\right)}. \quad (1)$$

142 The proof of the lemma is provided in Appendix.
 143 We mention that, given X and the probability of equality,
 144 some optimization problems such as optimizing $H(Y)$ and
 145 minimizing $H(X, Y)$ were solved by Prelov in [18] and [19].

146 As for the zero-error capacity region, Dueck [20, Sec. 2]
 147 established a characterization for a class of discrete memo-
 148 ryless multiple-access channels including the two-user union
 149 channel with complete feedback. However, pinning down
 150 the precise value of $R(\mathcal{O}_f)$ is still an open problem. For
 151 $q = 2$, the best lower bound $R(\mathcal{O}_f) \geq 0.78974$ was
 152 proved by Belokopytov [21] based on Dueck's characteriza-
 153 tion. Although Dueck's characterization shows that there
 154 exist good zero-error codes, it does not provide a way of
 155 constructing the best codes explicitly. If we use the scheme
 156 suggested by the proof of Dueck's theorem and generate a
 157 code at random with the appropriate distribution, the code
 158 constructed is likely to be good. However, without some
 159 structure in the code, it is computationally very difficult to
 160 decode. Hence the theorem does not provide a practical coding
 161 scheme.

162 In the context of group testing or a search problem
 163 on graphs, the “Fibonacci algorithm” by Christen [22]
 164 and Aigner [23] gives an $(F_{n+1}, F_n, n, 0)$ code explicitly,
 165 where F_n is the n th Fibonacci number, which implies that
 166 $R(\mathcal{O}_f) \geq \log_2 \phi = 0.69424$, where $\phi = 1.61834$ is the
 167 golden ratio. Later, the Fibonacci code was rediscovered

2 Salehi [16, Sec. III-D] showed that the channel capacity region is retained when the cardinality of U , denoted by $|U|$, is restricted to $\binom{q+1}{2}$. This bound on $|U|$ was also mentioned in [14]. However, [15] only referred to a slightly weaker bound $|U| \leq \binom{q+1}{2} + 2$.

TABLE I
SUMMARY OF BOUNDS ON THE AVERAGE CAPACITIES
OF TWO-USER UNION CHANNELS

q	$R(\mathcal{O})$	$R(\mathcal{E})$	$R(\mathcal{O}_f)$	$R(\mathcal{E}_f)$
2	[0.65891, 0.75]	0.75	[0.78974, 0.79113]	0.79113
3	[0.69281, 0.78969]	0.78969	[0.81071, 0.81510]	0.81510
4	[0.71256, 0.8125]	0.8125	[0.82946, 0.83044]	0.83044
5	[0.72292, 0.82773]	0.82773	[0.84123, 0.84130]	0.84130
6	[0.72914, 0.83881]	0.83881	[0.84953, 0.84959]	0.84959

168 by Zhang *et al.* [24, Th. 1], and was refined [24, Th. 2] to
169 achieve $R(\mathcal{O}_f) \geq \log_2 \phi' = 0.71662$, where $\phi' = 1.64333$ is
170 the real root of $x^{11} = x^{10} + x^9 + 5$. Using the language of decision
171 trees, Gargano *et al.* [25] constructs a $(32, 32, 7, 0)$ code
172 in an attempt to improve the Fibonacci code. Before our work,
173 the best construction is a $(2^{235n+61}, 2^{235n+61}, 312n + 123, 0)$
174 code, for every $n \in \mathbb{N}$, due to Belokopytov and Luzgin [26],
175 achieving $R(\mathcal{O}_f) \geq 235/312 = 0.75321$. We present in
176 Section III a practical communication scheme which achieves
177 a near-optimal zero-error capacity for all q . For $q = 2$, our
178 scheme achieves $R(\mathcal{O}_f) \geq 0.77291$. For $q \geq 3$, our scheme
179 is new and provides a lower bound that is close to the current
180 upper bound.

181 We summarize the known results in Table I for $q \leq 6$ with
182 our contribution in bold. We conclude in Section IV with some
183 open problems.

184 II. CHANNEL CAPACITY WITH COMPLETE FEEDBACK

185 Hereafter $\log x$ stands for the base- q logarithm of x .
186 We define the entropy functions

$$187 H(x_1, \dots, x_k) := -\sum x_i \log x_i, \text{ for } x_i \geq 0 \text{ and } \sum x_i = 1.$$

188 and we abbreviate

$$189 H\left(\underbrace{\frac{x_1}{r_1}, \dots, \frac{x_1}{r_1}}_{r_1}, \dots, \underbrace{\frac{x_k}{r_k}, \dots, \frac{x_k}{r_k}}_{r_k}\right)$$

190 by $H(x_1, \dots, x_k; r_1, \dots, r_k)$.

191 Under this notation, the function $F: [0, 1] \rightarrow \mathbb{R}$ defined in
192 Lemma 2 can be written as

$$193 F(\theta) = 2H(\alpha, 1-\alpha; 1, q-1), \text{ for } \theta \in [1/q, 1], \quad (2)$$

194 where $\alpha = \alpha(\theta)$, as in (1), is the larger root of the quadratic
195 equation

$$196 (1-\alpha)^2 = (q-1)(\theta - \alpha^2) \\ 197 \text{or equivalently } q\alpha^2 - 2\alpha + 1 = (q-1)\theta. \quad (3)$$

198 To express the average channel capacity $R(\mathcal{E}_f)$, we need the
199 concave envelope of F and another function $G: [0, 1] \rightarrow \mathbb{R}$.

200 *Definition 1:* The concave envelope of a continuous function
201 $F: I \rightarrow \mathbb{R}$ on a closed interval I , denoted by \hat{F} , is the
202 lowest-valued concave function that overestimates or equals F
203 over I . It follows from the strengthened Carathéodory theorem

204 by Fenchel and Eggleston [27, Th. 18] that for every $\theta \in I$,
205 \hat{F} is given by

$$206 \hat{F}(\theta) = \max \{p_1 F(\theta_1) + p_2 F(\theta_2) : 0 \leq p_1, p_2 \leq 1, \\ 207 \theta_1, \theta_2 \in I, p_1 + p_2 = 1, p_1\theta_1 + p_2\theta_2 = \theta\}.$$

208 *Lemma 3:* The function $G: [0, 1] \rightarrow \mathbb{R}$ defined by

$$209 G(\theta) := H\left(\theta, 1-\theta; q, \binom{q}{2}\right) \\ 210 = H(\theta, 1-\theta) + \theta + (1-\theta) \log \binom{q}{2} \quad (4)$$

211 is concave and it attains its maximum $\log \binom{q+1}{2}$ at $2/(q+1)$.

212 *Proof:* Since $H(\theta, 1-\theta)$ is concave and $G(\theta) - H(\theta, 1-\theta)$ is a linear function of θ , we conclude that $G(\theta)$
213 is also concave. Taking the derivative of G

$$214 G'(\theta) = \log\left(\frac{1-\theta}{\theta}\right) + 1 - \log\binom{q}{2} \quad (215)$$

216 and solving $G'(\theta) = 0$ yields the maximum point
217 $\theta = 2/(q+1)$. \square

218 *Theorem 4:* The average channel capacity $R(\mathcal{E}_f)$ of the
219 two-users union channel using input alphabet $[q]$ with complete
220 feedback is given by

$$221 R := \frac{1}{2} \max \left\{ \min \left(\hat{F}(\theta), G(\theta) \right) : \theta \in I \right\}, \quad (5)$$

222 where F and G are defined by (2) and (4), and $I :=$
223 $[1/q, 2/(q+1)]$. Moreover, the concave envelope of F on I
224 is given by

$$225 \hat{F}(\theta) = \begin{cases} F(\theta) & \text{when } q = 2; \\ 2 - \frac{2(q-1)\log(q-1)}{q-2} \left(\theta - \frac{1}{q}\right) & \text{when } q \geq 3, \end{cases}$$

226 and R can thus be simplified to $\frac{1}{2}\hat{F}(\theta)$, where θ is

- 227 1) the solution of $\hat{F}(\theta) = G(\theta)$ in I , when $q = 2, 3, 4$;
- 228 2) simply $\frac{2}{q+1}$, when $q \geq 5$.

229 *Proof:* We first choose random variables X_1, X_2, U in
230 Theorem 1 to demonstrate that (R, R) is in the channel capac-
231 ity region \mathcal{E}_f . Let $\theta \in I$ be a maximizer of $\min(\hat{F}(\theta), G(\theta))$
232 in (5), that is, R is the minimum of $\frac{1}{2}\hat{F}(\theta)$ and $\frac{1}{2}G(\theta)$.
233 By the definition of concave envelope, there are $p_1, p_2 \in$
234 $[0, 1], \theta_1, \theta_2 \in [1/q, 1]$ such that $p_1 + p_2 = 1, p_1\theta_1 + p_2\theta_2 = \theta$
235 and $p_1F(\theta_1) + p_2F(\theta_2) = \hat{F}(\theta) \leq R$. Choose the discrete
236 random variable U as follows: for every $u \in \{1, 2\}, v \in [q]$,
237 $U = (u, v)$ with probability p_u/q . Given $U = (u, v) \in$
238 $\{1, 2\} \times [q]$, we choose two conditionally independent random
239 variables X_1, X_2 :

$$240 X_1, X_2 = \begin{cases} v & \text{with probability } \alpha(\theta_u), \\ v' & \text{with probability } \frac{1-\alpha(\theta_u)}{q-1}, \end{cases} \text{ for } v' \in [q] \setminus \{v\}.$$

241 Based on these choices of random variables, we obtain that
242 for every $v \in [q]$,

$$243 \Pr(X_1 = X_2 = v) \\ 244 = \sum_{u=1}^2 \frac{p_u}{q} \left(\alpha^2(\theta_u) + (q-1) \left(\frac{1-\alpha(\theta_u)}{q-1} \right)^2 \right) \\ 245 \stackrel{(3)}{=} \sum_{u=1}^2 \frac{p_u}{q} \theta_u = \frac{\theta}{q},$$

246 and for every $v_1 \neq v_2$,

$$\begin{aligned} 247 \quad & \Pr(X_1 = v_1, X_2 = v_2) \\ 248 \quad &= \sum_{u=1}^2 \frac{p_u}{q} \left(2\alpha(\theta_u) \frac{1 - \alpha(\theta_u)}{q-1} + (q-2) \left(\frac{1 - \alpha(\theta_u)}{q-1} \right)^2 \right) \\ 249 \quad &\stackrel{(3)}{=} \sum_{u=1}^2 \frac{p_u}{q} \frac{1 - \theta_u}{q-1} = \frac{1 - \theta}{q(q-1)}. \end{aligned}$$

250 Then according to Theorem 1, the channel capacity region \mathcal{E}_f
251 contains all (R_1, R_2) satisfying,

$$\begin{aligned} 252 \quad & R_i \leq H(X_i | U) \\ 253 \quad &= \sum_{u=1}^2 p_u H(\alpha(\theta_u), 1 - \alpha(\theta_u); 1, q-1) \\ 254 \quad &= \sum_{u=1}^2 p_u \frac{1}{2} F(\theta_u) = \frac{1}{2} \hat{F}(\theta), \quad \text{for } i \in \{1, 2\}, \\ 255 \quad & R_1 + R_2 \leq H(\{X_1, X_2\}) \\ 256 \quad &= H\left(\theta, 1 - \theta; q, \binom{q}{2}\right) = G(\theta). \end{aligned}$$

257 Clearly (R, R) satisfies these conditions, and so $(R, R) \in \mathcal{E}_f$.

258 Next we give a proof that \mathcal{E}_f is a subset of the half-space
259 $\{(R_1, R_2) : \frac{1}{2}(R_1 + R_2) \leq R\}$. Since the half-space is already
260 convex, from Theorem 1, it suffices to prove that

$$261 \quad \min(H(X | U) + H(Y | U), H(\{X_1, X_2\})) \leq 2R, \quad (6)$$

262 for every discrete random variable U and $[q]$ -valued random
263 variables X_1, X_2 that are conditionally independent given U .
264 Without loss of generality, we may assume that $U = u$ with
265 probability p_u . Set

$$\begin{aligned} 266 \quad & \theta_{\{v\}}^u := \Pr(\{X_1, X_2\} = \{v\} | U = u), \\ 267 \quad & \quad \text{for all } u \text{ and } v \in [q], \\ 268 \quad & \theta_{\{v_1, v_2\}}^u := \Pr(\{X_1, X_2\} = \{v_1, v_2\} | U = u), \\ 269 \quad & \quad \text{for all } u \text{ and } v_1, v_2 \in [q], v_1 \neq v_2, \\ 270 \quad & \theta^u := \Pr(X_1 = X_2 | U = u), \\ 271 \quad & \theta := \Pr(X_1 = X_2) = \sum_u p_u \theta_u. \end{aligned}$$

272 On the one hand, as X_1 and X_2 are conditionally independent
273 given U , by the definition of F in Lemma 2, we have

$$\begin{aligned} 274 \quad & H(X_1 | U) + H(X_2 | U) = \sum_u p_u H(X_1, X_2 | U = u) \\ 275 \quad & \leq \sum_u p_u F(\theta^u) \leq \hat{F}(\theta). \end{aligned} \quad (7)$$

276 On the other hand, because $x \mapsto -x \log x$ is concave,
277 we obtain from Jensen's inequality that

$$\begin{aligned} 278 \quad & H(\{X_1, X_2\}) \\ 279 \quad &= H\left(\left(\sum_u p_u \theta_{\{v\}}^u\right)_{v=1}^q, \left(\sum_u p_u \theta_{\{v_1, v_2\}}^u\right)_{v_1 \neq v_2}\right) \\ 280 \quad &\leq H\left(\theta, 1 - \theta; q, \binom{q}{2}\right) = G(\theta). \end{aligned} \quad (8)$$

Combining (7) and (8), the left hand side of (6) is at most

$$\begin{aligned} 282 \quad & \min(\hat{F}(\theta), G(\theta)) \\ 283 \quad &\leq \max\{\min(\hat{F}(\theta), G(\theta)) : \theta \in [0, 1]\}. \end{aligned} \quad (9)$$

From Lemma 2, F is increasing on $[0, 1/q]$ and decreasing on $[1/q, 1]$, so is its concave envelope \hat{F} . Combining with Lemma 3 which says that G is increasing on $[0, 2/(q+1)]$ and decreasing on $[2/(q+1), 1]$, we can find a maximizer of the right hand side of (9) in I . Therefore the right hand side of (9) equals $2R$.

By the unimodality of F , we know that \hat{F} restricted to $[1/q, 1]$ is the same as the concave envelope of $F_0 := F|_{[1/q, 1]}$, the explicit formula of which is given by (2). We are left to find a maximizer of $M: I \rightarrow \mathbb{R}$ defined by $M(\theta) := \min(\hat{F}_0(\theta), G(\theta))$. Since \hat{F}_0 is increasing on I , G is decreasing on I , and $\hat{F}_0(1/q) > G(1/q)$, the maximizer of M depends on which of $\hat{F}_0(2/(q+1))$ and $G(2/(q+1))$ is larger.

Case $q = 2$: Observe that F_0 is concave already, and so $\hat{F}_0 = F_0$. Since $F_0(2/(q+1)) < G(2/(q+1))$, the maximizer of M is the solution of the equation $F(\theta) = G(\theta)$ in I .

Case $q \geq 3$: Observe that F_0 has an inflection point $\theta^* \in (1/q, 1)$, and F_0 is convex on $[1/q, \theta^*]$ and concave on $[\theta^*, 1]$. Let $\theta' \in (\theta^*, 1)$ be the point such that the line through the point $(1/q, F(1/q))$ and the point $(\theta', F(\theta'))$ is above the graph of F . In fact θ' is the root of the equation

$$\frac{F(\theta') - F(1/q)}{\theta' - 1/q} = F'(\theta') \quad 308$$

in $(1/q, 1)$, which turns out to be $\theta' = \frac{1}{q} + \frac{(q-2)^2}{q(q-1)}$. Define the linear function $L: [1/q, \theta'] \rightarrow \mathbb{R}$ to be

$$\begin{aligned} 311 \quad & L(\theta) = F(1/q) + F'(\theta') \left(\theta - \frac{1}{q} \right) \\ 312 \quad &= 2 - \frac{2(q-1)\log(q-1)}{q-2} \left(\theta - \frac{1}{q} \right). \end{aligned}$$

Indeed the graph of L is the line segment connecting $(1/q, F(1/q))$ and $(\theta', F(\theta'))$, and $\hat{F}_0(\theta) = L(\theta)$ for all $\theta \in (1/q, \theta')$. As $\theta' \geq 2/(q+1)$, we obtain that

$$\begin{aligned} 316 \quad & \hat{F}_0\left(\frac{2}{q+1}\right) = L\left(\frac{2}{q+1}\right) = 2 - \frac{2(q-1)^2 \log(q-1)}{(q-2)q(q+1)}, \\ 317 \quad & G\left(\frac{2}{q+1}\right) = \log\left(\frac{q+1}{2}\right). \end{aligned}$$

It is enough to determine the sign of

$$\begin{aligned} 319 \quad & (\ln q) \left(\hat{F}_0\left(\frac{2}{q+1}\right) - G\left(\frac{2}{q+1}\right) \right) \\ 320 \quad &= \ln\left(\frac{2q}{q+1}\right) - \frac{2(q-1)^2}{(q-2)q(q+1)} \ln(q-1) =: \Delta(q). \end{aligned}$$

Compute directly $\Delta(3) = \ln\frac{3}{2} - \frac{2}{3} \ln 2 < 0$ and $\Delta(4) = \ln\frac{8}{5} - \frac{9}{20} \ln 3 < 0$, thus the maximizer of M is the solution

of the equation $L(\theta) = G(\theta)$ in I . In the cases $q \geq 5$, we estimate

$$\begin{aligned}\Delta(q) &= \ln 2 + \ln \left(1 - \frac{1}{q+1}\right) \\ &\quad - \left(1 + \frac{1}{q(q-2)}\right) \frac{2 \ln(q-1)}{q} \\ &\geq \ln 2 + \ln \left(1 - \frac{1}{6}\right) - \left(1 + \frac{1}{15}\right) \frac{2 \ln 4}{5} > 0,\end{aligned}$$

thus the maximizer of M is simply $\frac{2}{q+1}$ and the maximum of M is $G(2/(q+1))$. \square

Remark 2: Observe that $R(\mathcal{E}_f) \leq \frac{1}{2} \log \binom{q+1}{2}$ is a naive bound since the output alphabet of the two-user union channel has cardinality $\binom{q+1}{2}$. In [2, Sec. II], it was stated that equality holds in this naive bound for $q \geq 6$. It was then conjectured there that in Theorem 1 a random variable U of cardinality q readily gives $(R, R) \in \mathcal{E}_f$. However, the random variable U used in our proof of Theorem 4 has cardinality $2q$, and we do not see a way to reduce its cardinality.

III. ZERO-ERROR CAPACITY WITH COMPLETE FEEDBACK

In this section, we describe a zero-error communication scheme for the two-user union channel with complete feedback and show that it achieves a near-optimal rate pair $(R, R) \in \mathcal{O}_f$. This scheme, like that in [14, Sec. IV], partitions the uses of the channel into a large number $B + 1$ of blocks, each of length n except the last block. Suppose the message sets of the senders are both $[q]^{Bm}$, where $m \in [n]$ will be decided later, and let $w_1, w_2 \in [q]^{Bm}$ be the messages of the two senders.

To describe the communication scheme in each block, we represent the uncertainty of the receiver about the first bm digits of w_1 and w_2 at the end of block b by $U(b) \subseteq ([q] \times [q])^{bm}$. In other words, the receiver, at the end of block b , knows that

$$((w_{1i}, w_{2i}))_{i=1}^{bm} \in U(b).$$

The key idea of our communication scheme is to keep the uncertainty sets uniformly bounded in size.

Due to the feedback of the channel, the uncertainty set is common knowledge between the senders and the receiver. The initial uncertainty set $U(0) := \emptyset$. In addition, we assume for a moment that

at the end of block b each sender knows

the first bm digits of the other message. (10)

This assumption will be shown below to hold by induction on the blocks.

A. Indexing

At the start of block $b + 1$, the senders and the receiver index the elements in $U(b)$ by

$$S := \{(s_1, \dots, s_n) : s_k \in \{\ast\} \cup [q] \text{ such that } |\{k : s_k = \ast\}| = m\}.$$

We shall choose $m \in [n]$ carefully in Theorem 5 so that $|U(b)| \leq |S| = \binom{n}{m} q^{n-m}$. The method of indexing can be agreed beforehand between the senders and the receiver. For example, they can order both $U(b)$ and S lexicographically, and index the elements in the ordered set $U(b)$ by the first $|U(b)|$ elements in S .

B. Encoding

During block $b + 1$, according to the inductive assumption (10) of our scheme, both senders know $((w_{1i}, w_{2i}))_{i=1}^{bm}$ and its index $(s_1, \dots, s_n) \in S$. During the k th use of the channel in block $b + 1$, both senders simply send s_k if $s_k \in [q]$; and send $w_{1,bm+i}$ and $w_{2,bm+i}$ respectively if s_k is the i th star in (s_1, \dots, s_n) . In the latter case, based on the feedback, each sender learns the $(bm+i)$ th digit of the other message. Because there are a total of m stars in (s_1, \dots, s_n) , at the end of block $b + 1$, each sender learns m more digits of the other message, maintaining the inductive assumption (10) of the scheme.

C. Decoding

After n uses of the channel in block $b + 1$, the receiver has received $(y_1, \dots, y_n) \in (\mathcal{Y}_1 \cup \mathcal{Y}_2)^n$, where $\mathcal{Y}_i := \{y \subset [q] : |y| = i\}$. The receiver then enumerates (s_1, \dots, s_n) through the first $|U(b)|$ elements in S , and for each (s_1, \dots, s_n) compatible with the output, namely $y_k = \{s_k\}$ if $s_k \in [q]$, the receiver adds to $U(b + 1)$ all the $((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{(b+1)m} \in ([q] \times [q])^{(b+1)m}$ such that

- 1) $((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{bm} \in U(b)$ is indexed by (s_1, \dots, s_n) ; and
- 2) $\{\hat{w}_{1,bm+i}, \hat{w}_{2,bm+i}\} = y_{k_i}$ for all $i \in [m]$, where k_1, \dots, k_m are the indices of s_k that are stars.

The updated uncertainty set $U(b + 1)$ will be shown in Theorem 5 to be bounded by $|S|$ in size.

After B blocks of uses of the channel, the receiver obtains the uncertainty set $U(B)$, and both senders know (w_1, w_2) and its index $(s_1, \dots, s_n) \in S$ as a member of $U(B)$. Finally, in the last block $B + 1$, the senders simply communicate the index (s_1, \dots, s_n) through $\lceil \log |S| \rceil \leq n - m + \lceil \log \binom{n}{m} \rceil$ uses of the channel.

Theorem 5: If $n/2 \leq m \leq n$ and

$$\binom{2n-2m}{n-m} 2^{2m-n} \leq \binom{n}{m} q^{n-m}, \quad (11)$$

then the communication scheme described above allows the receiver to recover the messages $w_1, w_2 \in [q]^{Bm}$ from the senders without errors through $\leq Bn + n - m + \lceil \log \binom{n}{m} \rceil$ uses of the two-user union channel with complete feedback. In particular, $R(\mathcal{O}_f) \geq R$, where R is the solution of $H_b(\alpha) + (1-\alpha) \log_2 q = 1$ in $(1/2, 1]$, and H_b is the binary entropy function.

Proof: It suffices to show that $|U(b)| \leq |S|$ for all $b = 0, 1, \dots, B$. The base case is evident as $U(0)$ consists of the empty sequence. For the inductive step, assume that $|U(b)| \leq |S|$. During the $(b + 1)$ st block, the receiver has received $(y_1, \dots, y_n) \in (\mathcal{Y}_1 \cup \mathcal{Y}_2)^n$.

We shall estimate the size of the uncertainty set $U(b + 1)$ at the end of the $(b + 1)$ st block. Suppose that $((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{bm} \in U(b)$ is indexed by (s_1, \dots, s_n) . Recall that if $s_k \in$

[q], then $y_k = \{s_k\}$; otherwise s_k is the i th star and $\{\hat{w}_{1,bm+i}, \hat{w}_{2,bm+i}\} = y_k$. In other words, only when $y_k = \{s_k\}$ for every k such that $s_k \in [q]$, the uncertainty set $U(b+1)$ would include the following $\prod_{i=1}^m |y_{k_i}|$ elements

$\{((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{(b+1)m}\}$ such that

$$\{\hat{w}_{1,bm+i}, \hat{w}_{2,bm+i}\} = y_{k_i} \text{ for all } i \in [m],$$

where k_1, \dots, k_m are the indices of s_k that are stars. Suppose $L := \{k : y_k \in \mathcal{Y}_2\}$ and $\ell := |L|$. A $(s_1, \dots, s_n) \in S$ compatible with (y_1, \dots, y_n) must have stars on coordinates indexed by L and choose from the rest $n - \ell$ positions an additional $m - \ell$ stars. This (s_1, \dots, s_n) , if it indexes an element in $U(b)$, will contribute at most 2^ℓ elements to $U(b+1)$. Therefore, we can estimate $|U(b+1)| \leq \binom{n-\ell}{m-\ell} 2^\ell$.

We claim that this estimate $u_\ell := \binom{n-\ell}{m-\ell} 2^\ell$ reaches its maximum $\binom{2n-2m}{n-m} 2^{2m-n}$ at $\ell = 2m-n, 2m-n+1$. In fact, we compare

$$\frac{u_\ell}{u_{\ell+1}} = \frac{\binom{n-\ell}{m-\ell} 2^\ell}{\binom{n-\ell-1}{m-\ell-1} 2^{\ell+1}} = \frac{n-\ell}{2(m-\ell)},$$

which is less than 1 when $\ell < 2m-n$ and greater than 1 when $\ell > 2m-n$. Combining with (11), it is guaranteed that $|U(b+1)| \leq |S|$. This finishes the inductive step.

Finally we prove the lower bound on $R(\mathcal{O}_f)$. Fix $\alpha \in (1/2, 1]$ such that $H_b(\alpha) + (1-\alpha) \log_2 q > 1$. It is well known that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log_2 \binom{n}{\lceil \alpha n \rceil} = H_b(\alpha).$$

We can choose n sufficiently large and $m = \lceil \alpha n \rceil$ so that

$$\begin{aligned} \frac{1}{n} \log_2 \binom{n}{m} + \left(1 - \frac{m}{n}\right) \log_2 q &\geq 1 \\ \implies \binom{n}{m} q^{n-m} &\geq 2^n \geq \binom{2n-2m}{n-m} 2^{2m-n}. \end{aligned}$$

Our communication scheme provides a $(q^{Bm}, q^{Bm}, Bn+n-m+\lceil \log \binom{n}{m} \rceil, 0)$ code. Thus

$$R(\mathcal{O}_f) \geq \frac{Bm}{Bn+n-m+\lceil \log \binom{n}{m} \rceil} \rightarrow \frac{m}{n} \geq \alpha \text{ as } B \rightarrow \infty.$$

Note that $H_b(\alpha) + (1-\alpha) \log_2 q - 1$ is decreasing on $(1/2, 1]$ and has a unique root $R \in [0, 1]$. We can choose α arbitrarily close to the root R to show that $R(\mathcal{O}_f) \geq R$. \square

Remark 3: For $q = 2$, our communication scheme could theoretically achieve $R(\mathcal{O}_f) \geq 0.77291$. In practice, in order to achieve $R(\mathcal{O}_f) \geq 0.764$, we can choose $m = 13, n = 17, B = 1019$ for our scheme to obtain a $(2^{13247}, 2^{13247}, 17339, 0)$ code. During the encoding and decoding process, the senders and the receiver need to keep track of up to 35840 binary numbers of length 13247 in the uncertainty set, which takes up 59.35 megabytes of memory for storage.

Remark 4: For $q \geq 5$, a naive upper bound on $R(\mathcal{O}_f)$ is $R(\mathcal{E}_f) = \frac{1}{2} \log \binom{q+1}{2} = 1 - \frac{1}{2 \log_2 q} + O(1/q)$. Notice that when $\alpha = 1 - \frac{1}{\log_2 q}$, $H_b(\alpha) + (1-\alpha) \log_2 q > (1-\alpha) \log_2 q = 1$. The proof of Theorem 5 implies that $R(\mathcal{O}_f) \geq 1 - \frac{1}{\log_2 q}$. Then

gap between the upper bound and the lower bound on $R(\mathcal{O}_f)$ is about $\frac{1}{2 \log_2 q}$.

IV. OPEN PROBLEMS

At the moment, the channel capacity $R(\mathcal{E}_f)$ of the union channel with complete feedback is determined for every $q \geq 2$. The zero-error capacity $R(\mathcal{O}_f)$ however is yet to be determined for any q . Naturally, the first step for future research is to determine $R(\mathcal{O}_f)$ for $q = 2$. Based on the characterization of \mathcal{O}_f by Dueck [20], our numerical experiments suggest that that for the binary case the lower bound on $R(\mathcal{O}_f)$ proved by Belokopytov [21] is tight.

Conjecture A: The average zero-error capacity of the two-user binary adder channel, that is the union channel with $q = 2$, with complete feedback is equal to 0.78974.

An inspection of Table I reveals that $R(\mathcal{E}) < R(\mathcal{O}_f)$ for $q \leq 6$. However, for $q \geq 14$, the lower bound on $R(\mathcal{O}_f)$ in Theorem 5 is less than $R(\mathcal{E})$. We speculate that our lower bound on $R(\mathcal{O}_f)$ can be improved for every $q \geq 2$.

Conjecture B: For all $q \geq 2$, the average channel capacity $R(\mathcal{E})$ of the two-user union channel without feedback is strictly less than the average zero-error capacity $R(\mathcal{O}_f)$ of the two-user union channel with complete feedback.

Conjecture B would establish the chain of inequalities $R(\mathcal{O}) \leq R(\mathcal{E}) < R(\mathcal{O}_f) \leq R(\mathcal{E}_f)$.

APPENDIX

Proof of Lemma 2: Let X_1 and X_2 be independent $[q]$ -valued random variables so that $\Pr(X_1 = X_2) = \theta$, and let $a_i := \Pr(X_1 = i)$ and $b_i := \Pr(X_2 = i)$ for every $i \in [q]$. Then $F(\theta)$ is

$$\text{the maximum of } H(a_1, \dots, a_q) + H(b_1, \dots, b_q), \quad (12a)$$

$$\text{subject to } \sum a_i = 1, \sum b_i = 1, \sum a_i b_i = \theta, \quad a_i, b_i \geq 0. \quad (12b)$$

Clearly $F(\theta)$ is continuous with respect to θ .

We first prove the unimodality of F , that is, for any $\theta \in [0, 1]$, if θ' is between θ and $1/\theta$, then $F(\theta') \geq F(\theta)$. Let $\mathbf{a} = (a_1, \dots, a_q), \mathbf{b} = (b_1, \dots, b_q)$ be a maximizer of the optimization problem (12). Let $\mathbf{j} = (1/q, \dots, 1/q) \in \mathbb{R}^q$, and consider $\mathbf{a}' = (1-t)\mathbf{a} + t\mathbf{j}$ and $\mathbf{b}' = (1-t)\mathbf{b} + t\mathbf{j}$, where $t \in [0, 1]$ is a solution of

$$\begin{aligned} \theta' &= \mathbf{a}' \cdot \mathbf{b}' = ((1-t)\mathbf{a} + t\mathbf{j}) \cdot ((1-t)\mathbf{b} + t\mathbf{j}) \\ &= (1-t)^2 \theta + \frac{(2-t)t}{q}. \end{aligned} \quad (13)$$

The right hand side of (13) equals θ and $1/q$ when $t = 0, 1$ respectively. By the intermediate value theorem, (13) has a solution in $[0, 1]$ and t is well-defined. Note that \mathbf{a}', \mathbf{b}' satisfy the constraint (12b). By the concavity of the entropy function, we have

$$\begin{aligned} F(\theta') &\geq H(\mathbf{a}') + H(\mathbf{b}') \geq (1-t)H(\mathbf{a}) + tH(\mathbf{j}) \\ &\quad + (1-t)H(\mathbf{b}) + tH(\mathbf{j}) \geq H(\mathbf{a}) + H(\mathbf{b}) = F(\theta). \end{aligned}$$

From this point forward, we assume that $\theta \in (1/q, 1]$ is fixed. We shall repeatedly add constraints to the optimization problem (12) without decreasing its maximum.

Given \mathbf{a}, \mathbf{b} satisfying the constraint (12b), consider the vectors \mathbf{a}', \mathbf{b}' whose coordinates are respectively the ones of \mathbf{a}, \mathbf{b} sorted in non-decreasing order. The rearrangement inequality says

$$\theta' := \left(\frac{\mathbf{a}' + \mathbf{b}'}{2} \right) \cdot \left(\frac{\mathbf{a}' + \mathbf{b}'}{2} \right) \geq \mathbf{a}' \cdot \mathbf{b}' \geq \mathbf{a} \cdot \mathbf{b} = \theta.$$

As θ is between $1/q$ and θ' , again we have $\mathbf{a}'' = \mathbf{b}'' = (1-t)\frac{\mathbf{a}'+\mathbf{b}'}{2} + t\mathbf{j}$ such that $\mathbf{a}'' \cdot \mathbf{b}'' = \theta$ for some $t \in [0, 1]$. By the concavity of the entropy function, we have

$$\begin{aligned} H(\mathbf{a}'') + H(\mathbf{b}'') &\geq 2 \left((1-t)H\left(\frac{\mathbf{a}'+\mathbf{b}'}{2}\right) + tH(\mathbf{j}) \right) \\ &\geq 2H\left(\frac{\mathbf{a}'+\mathbf{b}'}{2}\right) \geq H(\mathbf{a}') + H(\mathbf{b}') \\ &= H(\mathbf{a}) + H(\mathbf{b}). \end{aligned}$$

We come to the conclusion that without loss of generality we may assume $\mathbf{a} = \mathbf{b}$ in (12). The optimization problem is then equivalent to

$$\text{Maximize: } \frac{2}{\ln q} \left(- \sum_{i=1}^q x_i \ln x_i \right), \quad (14a)$$

$$\text{Subject to: } \sum_{i=1}^q x_i = 1, \quad \sum_{i=1}^q x_i^2 = \theta, \quad x_i \geq 0 \quad \text{for all } i \in [q]. \quad (14b)$$

Consider the Lagrangian

$$\begin{aligned} \mathcal{L}(\mathbf{x}, \lambda_1, \lambda_2) &:= - \sum_{i=1}^q x_i \ln x_i \\ &\quad + \lambda_1 \left(\sum_{i=1}^q x_i - 1 \right) + \lambda_2 \left(\sum_{i=1}^q x_i^2 - \theta \right). \end{aligned}$$

The method of Lagrange multipliers gives necessary conditions for the maximizers:

$$\frac{\partial \mathcal{L}}{\partial x_i} = -\ln x_i - 1 + \lambda_1 + 2\lambda_2 x_i = 0, \quad \text{for all } i \in [q].$$

In other words, given λ_1, λ_2 , each coordinate of a maximizer \mathbf{x} is a solution of the equation $-\ln x_i - 1 + \lambda_1 + 2\lambda_2 x_i = 0$. Since $x \mapsto -\ln x - 1 + \lambda_1 + 2\lambda_2 x$ is convex, there are at most two solutions. Without loss of generality, we can add to (14b) the constraints that

$$x_1 = \dots = x_r = a, \quad x_{r+1} = \dots = x_q = b,$$

for some $r \in [q-1]$, and $a \geq b \geq 0$. We have thus reduced the optimization problem (14a) to

$$\text{Maximize: } \frac{2}{\ln q} (-ra \ln a - sb \ln b), \quad (15a)$$

$$\text{Subject to: } ra + sb = 1, \quad ra^2 + sb^2 = \theta, \quad (15b)$$

$$r + s = q, \quad r, s \in [q], \quad a \geq b \geq 0. \quad (15c)$$

Given r, s such that $r + s = q$, we can solve from $ra + sb = 1, ra^2 + sb^2 = \theta$ for a, b :

$$\begin{aligned} (a, b) &= \left(\frac{1}{q} + \frac{\sqrt{rsp}}{qr}, \frac{1}{q} - \frac{\sqrt{rsp}}{qs} \right) \\ &\quad \text{or } \left(\frac{1}{q} - \frac{\sqrt{rsp}}{qr}, \frac{1}{q} + \frac{\sqrt{rsp}}{qs} \right). \end{aligned}$$

where $p := \theta q - 1 \in (0, q-1]$. Because $a \geq b$, we discard the second solution of (a, b) . Given the parameter $t := r/q \in [1/q, 1-1/q]$, the variables $r = qt, s = q - qt$ and

$$a = \frac{1}{q} \left(1 + \sqrt{\frac{1-t}{t} p} \right), \quad b = \frac{1}{q} \left(1 - \sqrt{\frac{t}{1-t} p} \right)$$

can be seen as functions of t , so can the objective function $v(t) := -ra \ln a - sb \ln b$. The derivative of v is

$$v' = -q(a \ln a + ta'(\ln a + 1) - b \ln b + (1-t)b'(\ln b + 1)).$$

The implicit differentiation of $ra + sb = 1, ra^2 + sb^2 = \theta$ yields

$$q(a + ta' - b + (1-t)b') = 0,$$

$$a^2 + 2taa' - b^2 + 2(1-t)bb' = 0.$$

which can be viewed as a system of linear equations of a', b' . Using $a > 1/q > b$, we deduce that

$$a' = -\frac{a-b}{2t}, \quad b' = -\frac{a-b}{2(1-t)}.$$

and we simplify v' as follows:

$$v' = -q(a \ln a - \frac{1}{2}(a-b)(\ln a + 1))$$

$$- b \ln b - \frac{1}{2}(a-b)(\ln b + 1))$$

$$= -q \left(\frac{a+b}{2} \ln \left(\frac{a}{b} \right) + b - a \right)$$

$$= qb \left(-\frac{1}{2}(r+1) \ln r - 1 + r \right),$$

where $r = a/b > 1$. One can check that $-\frac{1}{2}(r+1) \ln r - 1 + r < 0$ for $r > 1$, and so $v(t)$ is decreasing. In other words, the objective function v attains its maximum at $t = 1/q$ or equivalently the maximizer of (15) is given by $r = 1, s = q-1, a = \alpha, b = \frac{1-\alpha}{q-1}$, which leads to $F(\theta) = 2H(\alpha, 1-\alpha; 1, q-1)$. \square

ACKNOWLEDGEMENTS

We thank the referees for the suggestions which improved both the exposition of the paper and the clarity of the proofs.

REFERENCES

- [1] S. C. Chang and J. K. Wolf, "On the T -user M -frequency noiseless multiple-access channel with and without intensity information," *IEEE Trans. Inf. Theory*, vol. IT-27, no. 1, pp. 41–48, Jan. 1981, doi: [10.1109/TIT.1981.1056304](https://doi.org/10.1109/TIT.1981.1056304).
- [2] A. Vinck, W. Hoeks, and K. A. Post, "On the capacity of the two-user M -ary multiple-access channel with feedback (Corresp.)," *IEEE Trans. Inf. Theory*, vol. IT-31, no. 4, pp. 540–543, Jul. 1985, doi: [10.1109/TIT.1985.1057066](https://doi.org/10.1109/TIT.1985.1057066).
- [3] T. M. Cover and J. A. Thomas, *Elements of Information Theory*, 2nd ed. Hoboken, NJ, USA: Wiley, 2006.
- [4] R. Ahlswede, "Combination methods models," in *Rudolf Ahlswede's Lectures on Information Theory* (Foundations in Signal Processing, Communications and Networking), vol. 13. A. Ahlswede, I. Althöfer, C. Deppe, and U. Tamm. Eds. Cham, Switzerland: Springer, 2018, doi: [10.1007/978-3-319-53139-7](https://doi.org/10.1007/978-3-319-53139-7).
- [5] R. Ahlswede, "Multi-way communication channels," in *Proc. 2nd Int. Symp. Inf. Theory*. Tsahkadsor, Armenia, 1973, pp. 23–51.
- [6] H. H.-J. Liao, "Multiple access channels," Ph.D. dissertation, Dept. Elect. Eng., Univ. Hawaii, Honolulu, HI, USA, Sep. 1972.

- 605 [7] M. Mattas and P. R. J. Ostergard, "A new bound for the zero-
606 error capacity region of the two-user binary adder channel," *IEEE
607 Trans. Inf. Theory*, vol. 51, no. 9, pp. 3289–3291, Sep. 2005,
608 doi: [10.1109/TIT.2005.853309](https://doi.org/10.1109/TIT.2005.853309).
- 609 [8] B. Lindström, "Determination of two vectors from the sum," *J. Combinat. Theory*, vol. 6, no. 4, pp. 402–407, May 1969.
- 610 [9] G. Cohen, S. Litsyn, and G. Zémor, "Binary B_2 -sequences: A new
611 upper bound," *J. Combinat. Theory, Ser. A*, vol. 94, no. 1, pp. 152–155,
612 Apr. 2001. [Online]. Available: <http://www.sciencedirect.com/science/article/pii/S0097316500931273>
- 613 [10] F. Gao and G. Ge, "New bounds on separable codes for multimedia
614 fingerprinting," *IEEE Trans. Inf. Theory*, vol. 60, no. 9, pp. 5257–5262,
615 Sep. 2014, doi: [10.1109/TIT.2014.2331989](https://doi.org/10.1109/TIT.2014.2331989).
- 616 [11] S. R. Blackburn, "Probabilistic existence results for separable codes,"
617 *IEEE Trans. Inf. Theory*, vol. 61, no. 11, pp. 5822–5827, Nov. 2015,
618 doi: [10.1109/TIT.2015.2473848](https://doi.org/10.1109/TIT.2015.2473848) Nov.
- 619 [12] C. Shangguan, X. Wang, G. Ge, and Y. Miao, "New bounds for frame-
620 proof codes," *IEEE Trans. Inf. Theory*, vol. 63, no. 11, pp. 7247–7252,
621 Nov. 2017, doi: [10.1109/TIT.2017.2745619](https://doi.org/10.1109/TIT.2017.2745619).
- 622 [13] N. Gaarder and J. K. Wolf, "The capacity region of a multiple-access
623 discrete memoryless channel can increase with feedback (Corresp.),"
624 *IEEE Trans. Inf. Theory*, vol. IT-21, pp. 100–102, 1975.
- 625 [14] T. Cover and C. S. K. Leung, "An achievable rate region for the multiple-
626 access channel with feedback," *IEEE Trans. Inf. Theory*, vol. IT-27,
627 no. 3, pp. 292–298, May 1981, doi: [10.1109/TIT.1981.1056357](https://doi.org/10.1109/TIT.1981.1056357).
- 628 [15] F. M. J. Willems, "The feedback capacity region of a class of discrete
629 memoryless multiple access channels," *IEEE Trans. Inf. Theory*,
630 vol. IT-28, no. 1, pp. 93–95, Jun. 1982, doi: [10.1109/TIT.1982.1056437](https://doi.org/10.1109/TIT.1982.1056437).
- 631 [16] M. Salehi, "Cardinality bounds on auxiliary variables in multiple-
632 user theory via the method of Ahlswede and Korner," Dept. Statist.,
633 Stanford Univ., Stanford, CA, USA, Tech. Rep. 33, 1978.
- 634 [17] F. Willems, "On multiple access channels with feedback (Corresp.),"
635 *IEEE Trans. Inf. Theory*, vol. IT-30, no. 6, pp. 842–845, Nov. 1984.
- 636 [18] V. V. Prelov, "On one extreme value problem for entropy and error
637 probability," *Problems Inf. Transmiss.*, vol. 50, no. 3, pp. 203–216,
638 Jun. 2014, doi: [10.1134/S003294601403016](https://doi.org/10.1134/S003294601403016).
- 639 [19] V. V. Prelov, "On some extremal problems for mutual information and
640 entropy," *Problemy Peredachi Informatsii*, vol. 52, no. 4, pp. 3–13, 2016
641 doi: [10.1134/S0032946016040013](https://doi.org/10.1134/S0032946016040013).
- 642 [20] G. Dueck, "The zero error feedback capacity region of a certain class of
643 multiple-access channels," *Problems Control Inf. Theory*, vol. 14, no. 2,
644 pp. 89–103, 1985.
- 645 [21] A. Belokopytov, "On the zero error feedback capacity region of the
646 binary adder channel," *Problems Control Inf. Theory*, vol. 18, no. 2,
647 pp. 125–133, 1989.
- 648 [22] C. Christen, "A Fibonacci algorithm for the detection of two elements,"
649 Ph.D. dissertation, Dept. d'IRO, Univ. Montréal, Montréal, QC,
650 Canada, 1980.
- 651 [23] M. Aigner, "Search problems on graphs," *Discrete Appl. Math.*, vol. 14,
652 no. 3, pp. 215–230, Jul. 1986, doi: [10.1016/0166-218X\(86\)90026-0](https://doi.org/10.1016/0166-218X(86)90026-0).
- 653 [24] Z. Zhang, T. Berger, and J. Massey, "Some families of zero-error block
654 codes for the two-user binary adder channel with feedback," *IEEE Trans.
655 Inf. Theory*, vol. 33, no. 5, pp. 613–619, Sep. 1987.
- 656 [25] L. Gargano, V. Montour, G. Setaro, and U. Vaccaro, "An improved
657 algorithm for quantitative group testing," *Discrete Appl. Math.*,
658 vol. 36, no. 3, pp. 299–306, May 1992. [Online]. Available:
659 <http://www.sciencedirect.com/science/article/pii/0166218X9290260H>
- 660 [26] A. Y. Belokopytov and V. N. Luzgin, "Block transmission of information
661 in a summing multiple access channel with feedback," *Problems Inf.
662 Transmiss.*, vol. 23, no. 4, pp. 347–351, Apr. 1988.
- 663 [27] H. G. Eggleston, *Convexity* (Cambridge Tracts in Mathematics and
664 Mathematical Physics), vol. 47. New York, NY, USA: Cambridge Univ.
665 Press, 1958.
- 666
- Zilin Jiang** is an applied mathematics instructor at Massachusetts Institute of Technology. Up until July 2018, he was a postdoctoral fellow at Technion–Israel Institute of Technology. He graduated from Peking University in 2011 with a B.Sc. degree in Mathematics, and received his Ph.D. in Algorithm, Combinatorics and Optimization from Carnegie Mellon University in 2016.
- Nikita Polyanskii** was born in Russia in 1991. He received the M.Sc. degree in Mathematics and the Ph.D. degree in Mathematics from the Lomonosov Moscow State University, Moscow, Russia, in 2013 and 2016, respectively. During 2015–2017 he was a researcher at the Institute for Information Transmission Problems, Moscow, Russia, and a senior engineer at Huawei Technologies, Moscow, Russia. Since 2017 Nikita has been a postdoctoral researcher in the Department of Mathematics, Technion–Israel Institute of Technology, Haifa, Israel. Since 2018 he has been a research scientist in the Center for Computational and Data-Intensive Science and Engineering, Skolkovo Institute of Science and Technology, Moscow, Russia. His research interests include coding theory and its applications to communications, storage systems, and combinatorics.
- Ilya Vorobyev** received the M.Sc. degree in Mathematics and the Ph.D. degree in Mathematics from the Lomonosov Moscow State University in 2013 and 2017, respectively. In 2015–2017 he worked as a research engineer at Huawei R&D department in Moscow. He also was a researcher at the Institute for Information Transmission Problems, Moscow, in 2015–2017. Since 2017 Ilya has been a senior researcher in the Advanced Combinatorics and Complex Networks Lab, Moscow Institute of Physics and Technology. Since 2018 he has been a research scientist in the Center for Computational and Data-Intensive Science and Engineering, Skolkovo Institute of Science and Technology. His research interests include extremal combinatorics and coding theory.

On Capacities of the Two-User Union Channel With Complete Feedback

Zilin Jiang[✉], Nikita Polyanskii[✉], and Ilya Vorobyev[✉]

Abstract—The exact values of the optimal symmetric rate point in the Cover–Leung capacity region of the two-user union channel with complete feedback were determined by Willems when the size of the input alphabet is 2, and by Vinck *et al.* when the size is at least 6. We complete this line of research when the size of the input alphabet is 3, 4, or 5. The proof hinges on the technical lemma that concerns the maximal joint entropy of two independent random variables in terms of their probability of equality. For the zero-error capacity region, using superposition coding, we provide a practical near-optimal communication scheme which improves all the previous explicit constructions.

Index Terms—Union channel, feedback, channel capacity, zero-error capacity, entropy function.

I. INTRODUCTION

THE two-user union channel, first introduced in [1] and rediscovered in [2], is a discrete memoryless multiple-access channel¹: the channel takes symbols x_1, x_2 from the input alphabet $\mathcal{X} := [q] = \{1, 2, \dots, q\}$ given by two senders, and outputs the union $y = \{x_1, x_2\}$ from the output alphabet $\mathcal{Y} := \{y \subseteq [q] : |y| \in \{1, 2\}\}$. For the special case $q = 2$, the union channel coincides with the *two-user binary adder channel*.

Since a received $y \in \mathcal{Y}$ cannot be unambiguously decoded, the central problem in two-user communication theory is to coordinate the two senders to send simultaneously as much information as possible to a single receiver through n uses of the union channel.

Manuscript received April 23, 2018; revised September 22, 2018; accepted November 27, 2018. Z. Jiang was supported by the Israel Science Foundation (ISF) under Grant 1162/15 and Grant 936/16. N. Polyanskii was supported in part by ISF under Grant 1162/15 and Grant 326/17 and in part by the Russian Foundation for Basic Research (RFBR) under Grant 16-01-00440 A, Grant 18-07-01427 A, and Grant 18-31-00310 MOL_A. I. Vorobyev was supported by RFBR under Grant 16-01-00440 A, Grant 18-07-01427 A, and Grant 18-31-00361 MOL_A.

Z. Jiang was with the Technion–Israel Institute of Technology, Haifa 3200003, Israel. He is now with the Department of Mathematics, Massachusetts Institute of Technology, Cambridge, MA 02139 USA (e-mail: zilinj@mit.edu).

N. Polyanskii is with CDISE, Skolkovo Institute of Science and Technology, 121205 Moscow, Russia, and also with the Department of Mathematics, Technion–Israel Institute of Technology, Haifa 3200003, Israel (e-mail: nikita.polyansky@gmail.com).

I. Vorobyev is with CDISE, Skolkovo Institute of Science and Technology, 121205 Moscow, Russia, and also with the Moscow Institute of Physics and Technology, 141701 Dolgoprudny, Russia (e-mail: vorobyev.i.v@yandex.ru).

Communicated by M. Lentmaier, Associate Editor for Coding Theory.

Digital Object Identifier 10.1109/TIT.2018.2889250

¹The terminology from information theory used throughout the article is standard, and can be found in [3].

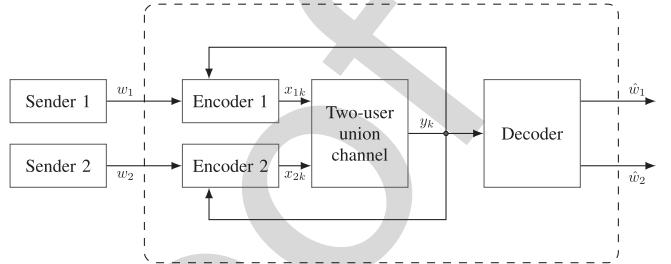


Fig. 1. Two-user union channel with complete feedback.

Let the message sets specified for the senders be of size M_1 and M_2 , and let $w_1 \in [M_1]$, $w_2 \in [M_2]$ be two messages chosen by the two senders beforehand. During the k th use of the channel, two functions e_{1k} and e_{2k} respectively encode w_1 and w_2 to two codewords $x_{1k} \in [q]$ and $x_{2k} \in [q]$. The union channel then takes x_{1k}, x_{2k} and outputs $y_k := \{x_{1k}, x_{2k}\} \in \mathcal{Y}$. The sequence of outputs $(y_k)_{k=1}^n$ is decoded by the receiver to the estimate (\hat{w}_1, \hat{w}_2) of (w_1, w_2) .

In such a communication scheme for the transmission of information from two senders to one receiver, the encoders at any moment might know some information about the signals received by the decoder prior to this moment. When the encoders know nothing but the message from their corresponding senders, we say that the channel is the *two-user union channel without feedback*. However, when the encoders also know all the previous outputs of the channel, namely every e_{1k} and e_{2k} depend not only on w_1 and w_2 respectively but also on $(y_i)_{i=1}^{k-1}$, we say that the channel is the *two-user union channel with complete feedback*. See [4, Sec. 4] for a broader view on coding for the multiple-access channel.

In this work, we mainly focus on capacities of the two-user union channel with complete feedback. An $(M_1, M_2, n, \varepsilon)$ code for the two-user union channel with complete feedback consists of a collection encoding functions and a decoding function such that the *probability of error*, defined by $\Pr((\hat{w}_1, \hat{w}_2) \neq (w_1, w_2))$ when (w_1, w_2) is drawn uniformly from $[M_1] \times [M_2]$, is at most ε . In particular, an $(M_1, M_2, n, 0)$ code could recover the messages without errors. The channel capacity region \mathcal{E}_f for the two-user union channel with complete feedback captures the rates at which the information can be transmitted over the channel for both users with arbitrarily small probability of error, whereas the zero-error capacity region \mathcal{O}_f represents the rates without error:

$$\mathcal{E}_f := \text{closure of } \{(R_1, R_2) : \exists \text{ a sequence of } ([q^{nR_1}], [q^{nR_2}], n, \varepsilon_n) \text{ codes s.t. } \varepsilon_n \rightarrow 0\},$$

63 $\mathcal{O}_f := \text{closure of } \{(R_1, R_2) : \exists \text{ a sequence of}$
 64 $(\lceil q^{nR_1} \rceil, \lceil q^{nR_2} \rceil, n, \varepsilon_n) \text{ codes s.t. } \varepsilon_n = 0 \text{ eventually}\}$.

65 In the absence of feedback, the channel capacity region \mathcal{E} and
 66 the zero-error capacity region \mathcal{O} are similarly defined for the
 67 two-user union channel.

68 For each of the above capacity regions, say \mathcal{C} , research has
 69 been devoted to the *average capacity*

$$70 R(\mathcal{C}) := \sup \left\{ \frac{1}{2}(R_1 + R_2) : (R_1, R_2) \in \mathcal{C} \right\},$$

71 which can be understood as the maximal rate per user at which
 72 the information can be transmitted. Because \mathcal{C} is convex and
 73 symmetric with respect to the line $R_1 = R_2$, the average
 74 capacity $R(\mathcal{C})$ can also be defined as $\sup \{R : (R, R) \in \mathcal{C}\}$.
 75 The point $(R(\mathcal{C}), R(\mathcal{C}))$ is known as the equal-rate point or
 76 the symmetric rate point in the existing literature.

77 The channel capacity region for a discrete memory-
 78 less multiple-access channel without feedback has been
 79 fully characterized by Alswede [5] and Liao [6]. For the
 80 two-user union channel without feedback, the average chan-
 81 nel capacity $R(\mathcal{E}) = 1 - \frac{q-1}{2q \log_2 q}$ has been determined
 82 by Chang and Wolf [1]. Much less is known for the aver-
 83 age zero-error capacity $R(\mathcal{O})$ of the two-user union chan-
 84 nel without feedback. For $q = 2$, there is no better
 85 upper bound other than the trivial $R(\mathcal{O}) \leq R(\mathcal{E}) =$
 86 0.75, while the current record lower bound is $R(\mathcal{O}) \geq$
 87 $\frac{1}{12} \log_2 240 = 0.65891$ due to Mattas and Östergård [7,
 88 Sec. III] obtained by computer searches. For $q \geq 3$, several
 89 constructions provided by Chang and Wolf [1, Sec. III] imply
 90 that $R(\mathcal{O}) \geq \frac{1}{4}(1 + \log_q(q^2 - q + 1))$ for all $q \geq 2$, $R(\mathcal{O}) \geq$
 91 $\log_q(\frac{1}{2}(q+1))$ for odd q and $R(\mathcal{O}) \geq \log_q(\frac{1}{2}\sqrt{q(q+2)})$ for
 92 even q . The variation where the senders are required to use
 93 the same encoding functions was studied in various context.
 94 We refer the readers to [8] for the best code construction when
 95 the size of the input alphabet is 2, to [9] for the connection
 96 with the binary B_2 -sequences, and to [10] for large input
 97 alphabet. The generalization, in which more than 2 users
 98 have access to the channel, was recently investigated in [11]
 99 and [12].

100 Gaarder and Wolf [13] demonstrated that feedback may
 101 increase the channel capacity region. They used the two-user
 102 binary adder channel as an example and developed a simple
 103 two-stage coding strategy. Using the concept of superposi-
 104 tion coding Cover and Leung [14] characterized a subset of
 105 the channel capacity region \mathcal{E}_f for the discrete memoryless
 106 multiple-access channels with complete feedback. This subset
 107 was later shown to be exactly \mathcal{E}_f by Willems [15] for the class
 108 of the channels where one of the inputs is determined by the
 109 other input and the output. Their results are paraphrased as
 110 the following theorem in the special case that the channel is
 111 the two-user union channel.

112 **Theorem 1 (Theorem 1 of Cover and Leung [14] and The-
 113 orem of Willems [15]):** The channel capacity region \mathcal{E}_f of the
 114 two-user union channel using input alphabet $[q]$ with complete
 115 feedback is the convex hull of all (R_1, R_2) satisfying

$$116 \quad 0 \leq R_1 \leq H(X_1 | U), \quad 0 \leq R_2 \leq H(X_2 | U),$$

$$117 \quad R_1 + R_2 \leq H(\{X_1, X_2\})$$

where U is a discrete random variable,² X_1, X_2 are two
 $[q]$ -valued random variables that are conditionally indepen-
 119 dent given U , and the entropy function H uses the base- q
 120 logarithm.

121 **Remark 1:** The entropy $H(\{X_1, X_2\})$ in Theorem 1 is the
 122 entropy of the random variable $Y := \{X_1, X_2\}$, not to be
 123 confused with the joint entropy $H(X_1, X_2)$.

124 For $q = 2$, Willems [17] later showed that $R(\mathcal{E}_f) =$
 125 0.79113; for $q \geq 6$, Vinck *et al.* [2] asserted that $R(\mathcal{E}_f) =$
 126 $\frac{1}{2} \log_q(\frac{q+1}{2})$. In Section II we complete this line of research on
 127 $R(\mathcal{E}_f)$ for all $q \geq 2$. The proof hinges on the following lemma
 128 about the maximum of the joint entropy of two independent
 129 discrete random variables in terms of their probability of
 130 equality.

131 **Lemma 2:** Given $q \geq 2$, for every $\theta \in [0, 1]$ let $F(\theta)$ be
 132 the maximum of the joint entropy $H(X_1, X_2)$ among all pairs
 133 of independent $[q]$ -valued random variables X_1, X_2 such that
 134 $\Pr(X_1 = X_2) = \theta$. The function $F: [0, 1] \rightarrow \mathbb{R}$ is continuous
 135 and it is increasing on $[0, 1/q]$ and decreasing on $[1/q, 1]$.
 136 Moreover,

$$137 F(\theta) = 2(-\alpha \log \alpha - (1-\alpha) \log(1-\alpha) + (1-\alpha) \log(q-1)), \quad \text{for } \theta \in [1/q, 1]$$

138 where the bijection $\alpha: [1/q, 1] \rightarrow [1/q, 1]$ is defined by

$$139 \alpha = \alpha(\theta) := \frac{1}{q} + \sqrt{\left(1 - \frac{1}{q}\right)\left(\theta - \frac{1}{q}\right)}. \quad (1)$$

140 The proof of the lemma is provided in Appendix.
 141 We mention that, given X and the probability of equality,
 142 some optimization problems such as optimizing $H(Y)$ and
 143 minimizing $H(X, Y)$ were solved by Prelov in [18] and [19].

144 As for the zero-error capacity region, Dueck [20, Sec. 2]
 145 established a characterization for a class of discrete memo-
 146 ryless multiple-access channels including the two-user union
 147 channel with complete feedback. However, pinning down
 148 the precise value of $R(\mathcal{O}_f)$ is still an open problem. For
 149 $q = 2$, the best lower bound $R(\mathcal{O}_f) \geq 0.78974$ was
 150 proved by Belokopytov [21] based on Dueck's characteriza-
 151 tion. Although Dueck's characterization shows that there
 152 exist good zero-error codes, it does not provide a way of
 153 constructing the best codes explicitly. If we use the scheme
 154 suggested by the proof of Dueck's theorem and generate a
 155 code at random with the appropriate distribution, the code
 156 constructed is likely to be good. However, without some
 157 structure in the code, it is computationally very difficult to
 158 decode. Hence the theorem does not provide a practical coding
 159 scheme.

160 In the context of group testing or a search problem
 161 on graphs, the “Fibonacci algorithm” by Christen [22]
 162 and Aigner [23] gives an $(F_{n+1}, F_n, n, 0)$ code explicitly,
 163 where F_n is the n th Fibonacci number, which implies that
 164 $R(\mathcal{O}_f) \geq \log_2 \phi = 0.69424$, where $\phi = 1.61834$ is the
 165 golden ratio. Later, the Fibonacci code was rediscovered

166 ²Salehi [16, Sec. III-D] showed that the channel capacity region is retained
 167 when the cardinality of U , denoted by $|U|$, is restricted to $(\frac{q-1}{2})$. This bound
 168 on $|U|$ was also mentioned in [14]. However, [15] only referred to a slightly
 169 weaker bound $|U| \leq (\frac{q+1}{2}) + 2$.

TABLE I
SUMMARY OF BOUNDS ON THE AVERAGE CAPACITIES
OF TWO-USER UNION CHANNELS

q	$R(\mathcal{O})$	$R(\mathcal{E})$	$R(\mathcal{O}_f)$	$R(\mathcal{E}_f)$
2	[0.65891, 0.75]	0.75	[0.78974, 0.79113]	0.79113
3	[0.69281, 0.78969]	0.78969	[0.81071, 0.81510]	0.81510
4	[0.71256, 0.8125]	0.8125	[0.82946, 0.83044]	0.83044
5	[0.72292, 0.82773]	0.82773	[0.84123, 0.84130]	0.84130
6	[0.72914, 0.83881]	0.83881	[0.84953, 0.84959]	0.84959

168 by Zhang *et al.* [24, Th. 1], and was refined [24, Th. 2] to
169 achieve $R(\mathcal{O}_f) \geq \log_2 \phi' = 0.71662$, where $\phi' = 1.64333$ is
170 the real root of $x^{11} = x^{10} + x^9 + 5$. Using the language of decision
171 trees, Gargano *et al.* [25] constructs a $(32, 32, 7, 0)$ code
172 in an attempt to improve the Fibonacci code. Before our work,
173 the best construction is a $(2^{235n+61}, 2^{235n+61}, 312n + 123, 0)$
174 code, for every $n \in \mathbb{N}$, due to Belokopytov and Luzgin [26],
175 achieving $R(\mathcal{O}_f) \geq 235/312 = 0.75321$. We present in
176 Section III a practical communication scheme which achieves
177 a near-optimal zero-error capacity for all q . For $q = 2$, our
178 scheme achieves $R(\mathcal{O}_f) \geq 0.77291$. For $q \geq 3$, our scheme
179 is new and provides a lower bound that is close to the current
180 upper bound.

181 We summarize the known results in Table I for $q \leq 6$ with
182 our contribution in bold. We conclude in Section IV with some
183 open problems.

184 II. CHANNEL CAPACITY WITH COMPLETE FEEDBACK

185 Hereafter $\log x$ stands for the base- q logarithm of x .
186 We define the entropy functions

$$187 H(x_1, \dots, x_k) := -\sum x_i \log x_i, \text{ for } x_i \geq 0 \text{ and } \sum x_i = 1.$$

188 and we abbreviate

$$189 H\left(\underbrace{\frac{x_1}{r_1}, \dots, \frac{x_1}{r_1}}_{r_1}, \dots, \underbrace{\frac{x_k}{r_k}, \dots, \frac{x_k}{r_k}}_{r_k}\right)$$

190 by $H(x_1, \dots, x_k; r_1, \dots, r_k)$.

191 Under this notation, the function $F: [0, 1] \rightarrow \mathbb{R}$ defined in
192 Lemma 2 can be written as

$$193 F(\theta) = 2H(\alpha, 1-\alpha; 1, q-1), \text{ for } \theta \in [1/q, 1], \quad (2)$$

194 where $\alpha = \alpha(\theta)$, as in (1), is the larger root of the quadratic
195 equation

$$196 (1-\alpha)^2 = (q-1)(\theta - \alpha^2) \\ 197 \text{or equivalently } q\alpha^2 - 2\alpha + 1 = (q-1)\theta. \quad (3)$$

198 To express the average channel capacity $R(\mathcal{E}_f)$, we need the
199 concave envelope of F and another function $G: [0, 1] \rightarrow \mathbb{R}$.

200 *Definition 1:* The concave envelope of a continuous function
201 $F: I \rightarrow \mathbb{R}$ on a closed interval I , denoted by \hat{F} , is the
202 lowest-valued concave function that overestimates or equals F
203 over I . It follows from the strengthened Carathéodory theorem

204 by Fenchel and Eggleston [27, Th. 18] that for every $\theta \in I$,
205 \hat{F} is given by

$$206 \hat{F}(\theta) = \max \{p_1 F(\theta_1) + p_2 F(\theta_2) : 0 \leq p_1, p_2 \leq 1, \\ 207 \theta_1, \theta_2 \in I, p_1 + p_2 = 1, p_1\theta_1 + p_2\theta_2 = \theta\}.$$

208 *Lemma 3:* The function $G: [0, 1] \rightarrow \mathbb{R}$ defined by

$$209 G(\theta) := H\left(\theta, 1-\theta; q, \binom{q}{2}\right) \\ 210 = H(\theta, 1-\theta) + \theta + (1-\theta) \log \binom{q}{2} \quad (4)$$

211 is concave and it attains its maximum $\log \binom{q+1}{2}$ at $2/(q+1)$.

212 *Proof:* Since $H(\theta, 1-\theta)$ is concave and $G(\theta) - H(\theta, 1-\theta)$ is a linear function of θ , we conclude that $G(\theta)$
213 is also concave. Taking the derivative of G

$$214 G'(\theta) = \log\left(\frac{1-\theta}{\theta}\right) + 1 - \log\binom{q}{2} \quad (215)$$

216 and solving $G'(\theta) = 0$ yields the maximum point
217 $\theta = 2/(q+1)$. \square

218 *Theorem 4:* The average channel capacity $R(\mathcal{E}_f)$ of the
219 two-users union channel using input alphabet $[q]$ with complete
220 feedback is given by

$$221 R := \frac{1}{2} \max \left\{ \min \left(\hat{F}(\theta), G(\theta) \right) : \theta \in I \right\}, \quad (5)$$

222 where F and G are defined by (2) and (4), and $I :=$
223 $[1/q, 2/(q+1)]$. Moreover, the concave envelope of F on I
224 is given by

$$225 \hat{F}(\theta) = \begin{cases} F(\theta) & \text{when } q = 2; \\ 2 - \frac{2(q-1)\log(q-1)}{q-2} \left(\theta - \frac{1}{q}\right) & \text{when } q \geq 3, \end{cases}$$

226 and R can thus be simplified to $\frac{1}{2}\hat{F}(\theta)$, where θ is

- 227 1) the solution of $\hat{F}(\theta) = G(\theta)$ in I , when $q = 2, 3, 4$;
- 228 2) simply $\frac{2}{q+1}$, when $q \geq 5$.

229 *Proof:* We first choose random variables X_1, X_2, U in
230 Theorem 1 to demonstrate that (R, R) is in the channel capac-
231 ity region \mathcal{E}_f . Let $\theta \in I$ be a maximizer of $\min(\hat{F}(\theta), G(\theta))$
232 in (5), that is, R is the minimum of $\frac{1}{2}\hat{F}(\theta)$ and $\frac{1}{2}G(\theta)$.
233 By the definition of concave envelope, there are $p_1, p_2 \in$
234 $[0, 1], \theta_1, \theta_2 \in [1/q, 1]$ such that $p_1 + p_2 = 1, p_1\theta_1 + p_2\theta_2 = \theta$
235 and $p_1F(\theta_1) + p_2F(\theta_2) = \hat{F}(\theta) \leq R$. Choose the discrete
236 random variable U as follows: for every $u \in \{1, 2\}, v \in [q]$,
237 $U = (u, v)$ with probability p_u/q . Given $U = (u, v) \in$
238 $\{1, 2\} \times [q]$, we choose two conditionally independent random
239 variables X_1, X_2 :

$$240 X_1, X_2 = \begin{cases} v & \text{with probability } \alpha(\theta_u), \\ v' & \text{with probability } \frac{1-\alpha(\theta_u)}{q-1}, \end{cases} \text{ for } v' \in [q] \setminus \{v\}.$$

241 Based on these choices of random variables, we obtain that
242 for every $v \in [q]$,

$$243 \Pr(X_1 = X_2 = v) \\ 244 = \sum_{u=1}^2 \frac{p_u}{q} \left(\alpha^2(\theta_u) + (q-1) \left(\frac{1-\alpha(\theta_u)}{q-1} \right)^2 \right) \\ 245 \stackrel{(3)}{=} \sum_{u=1}^2 \frac{p_u}{q} \theta_u = \frac{\theta}{q},$$

246 and for every $v_1 \neq v_2$,

$$\begin{aligned} 247 \quad & \Pr(X_1 = v_1, X_2 = v_2) \\ 248 \quad &= \sum_{u=1}^2 \frac{p_u}{q} \left(2\alpha(\theta_u) \frac{1 - \alpha(\theta_u)}{q-1} + (q-2) \left(\frac{1 - \alpha(\theta_u)}{q-1} \right)^2 \right) \\ 249 \quad &\stackrel{(3)}{=} \sum_{u=1}^2 \frac{p_u}{q} \frac{1 - \theta_u}{q-1} = \frac{1 - \theta}{q(q-1)}. \end{aligned}$$

250 Then according to Theorem 1, the channel capacity region \mathcal{E}_f
251 contains all (R_1, R_2) satisfying,

$$\begin{aligned} 252 \quad & R_i \leq H(X_i | U) \\ 253 \quad &= \sum_{u=1}^2 p_u H(\alpha(\theta_u), 1 - \alpha(\theta_u); 1, q-1) \\ 254 \quad &= \sum_{u=1}^2 p_u \frac{1}{2} F(\theta_u) = \frac{1}{2} \hat{F}(\theta), \quad \text{for } i \in \{1, 2\}, \\ 255 \quad & R_1 + R_2 \leq H(\{X_1, X_2\}) \\ 256 \quad &= H\left(\theta, 1 - \theta; q, \binom{q}{2}\right) = G(\theta). \end{aligned}$$

257 Clearly (R, R) satisfies these conditions, and so $(R, R) \in \mathcal{E}_f$.

258 Next we give a proof that \mathcal{E}_f is a subset of the half-space
259 $\{(R_1, R_2) : \frac{1}{2}(R_1 + R_2) \leq R\}$. Since the half-space is already
260 convex, from Theorem 1, it suffices to prove that

$$261 \quad \min(H(X | U) + H(Y | U), H(\{X_1, X_2\})) \leq 2R, \quad (6)$$

262 for every discrete random variable U and $[q]$ -valued random
263 variables X_1, X_2 that are conditionally independent given U .
264 Without loss of generality, we may assume that $U = u$ with
265 probability p_u . Set

$$\begin{aligned} 266 \quad & \theta_{\{v\}}^u := \Pr(\{X_1, X_2\} = \{v\} | U = u), \\ 267 \quad & \quad \text{for all } u \text{ and } v \in [q], \\ 268 \quad & \theta_{\{v_1, v_2\}}^u := \Pr(\{X_1, X_2\} = \{v_1, v_2\} | U = u), \\ 269 \quad & \quad \text{for all } u \text{ and } v_1, v_2 \in [q], v_1 \neq v_2, \\ 270 \quad & \theta^u := \Pr(X_1 = X_2 | U = u), \quad \text{for all } u, \\ 271 \quad & \theta := \Pr(X_1 = X_2) = \sum_u p_u \theta_u. \end{aligned}$$

272 On the one hand, as X_1 and X_2 are conditionally independent
273 given U , by the definition of F in Lemma 2, we have

$$\begin{aligned} 274 \quad & H(X_1 | U) + H(X_2 | U) = \sum_u p_u H(X_1, X_2 | U = u) \\ 275 \quad & \leq \sum_u p_u F(\theta^u) \leq \hat{F}(\theta). \quad (7) \end{aligned}$$

276 On the other hand, because $x \mapsto -x \log x$ is concave,
277 we obtain from Jensen's inequality that

$$\begin{aligned} 278 \quad & H(\{X_1, X_2\}) \\ 279 \quad &= H\left(\left(\sum_u p_u \theta_{\{v\}}^u\right)_{v=1}^q, \left(\sum_u p_u \theta_{\{v_1, v_2\}}^u\right)_{v_1 \neq v_2}\right) \\ 280 \quad &\leq H\left(\theta, 1 - \theta; q, \binom{q}{2}\right) = G(\theta). \quad (8) \end{aligned}$$

Combining (7) and (8), the left hand side of (6) is at most

$$\begin{aligned} 282 \quad & \min(\hat{F}(\theta), G(\theta)) \\ 283 \quad &\leq \max\{\min(\hat{F}(\theta), G(\theta)) : \theta \in [0, 1]\}. \end{aligned} \quad (9)$$

From Lemma 2, F is increasing on $[0, 1/q]$ and decreasing on $[1/q, 1]$, so is its concave envelope \hat{F} . Combining with Lemma 3 which says that G is increasing on $[0, 2/(q+1)]$ and decreasing on $[2/(q+1), 1]$, we can find a maximizer of the right hand side of (9) in I . Therefore the right hand side of (9) equals $2R$.

By the unimodality of F , we know that \hat{F} restricted to $[1/q, 1]$ is the same as the concave envelope of $F_0 := F|_{[1/q, 1]}$, the explicit formula of which is given by (2). We are left to find a maximizer of $M: I \rightarrow \mathbb{R}$ defined by $M(\theta) := \min(\hat{F}_0(\theta), G(\theta))$. Since \hat{F}_0 is increasing on I , G is decreasing on I , and $\hat{F}_0(1/q) > G(1/q)$, the maximizer of M depends on which of $\hat{F}_0(2/(q+1))$ and $G(2/(q+1))$ is larger.

Case $q = 2$: Observe that F_0 is concave already, and so $\hat{F}_0 = F_0$. Since $F_0(2/(q+1)) < G(2/(q+1))$, the maximizer of M is the solution of the equation $F(\theta) = G(\theta)$ in I .

Case $q \geq 3$: Observe that F_0 has an inflection point $\theta^* \in (1/q, 1)$, and F_0 is convex on $[1/q, \theta^*]$ and concave on $[\theta^*, 1]$. Let $\theta' \in (\theta^*, 1)$ be the point such that the line through the point $(1/q, F(1/q))$ and the point $(\theta', F(\theta'))$ is above the graph of F . In fact θ' is the root of the equation

$$\frac{F(\theta') - F(1/q)}{\theta' - 1/q} = F'(\theta') \quad 308$$

in $(1/q, 1)$, which turns out to be $\theta' = \frac{1}{q} + \frac{(q-2)^2}{q(q-1)}$. Define the linear function $L: [1/q, \theta'] \rightarrow \mathbb{R}$ to be

$$\begin{aligned} 311 \quad & L(\theta) = F(1/q) + F'(\theta') \left(\theta - \frac{1}{q} \right) \\ 312 \quad &= 2 - \frac{2(q-1)\log(q-1)}{q-2} \left(\theta - \frac{1}{q} \right). \end{aligned}$$

Indeed the graph of L is the line segment connecting $(1/q, F(1/q))$ and $(\theta', F(\theta'))$, and $\hat{F}_0(\theta) = L(\theta)$ for all $\theta \in (1/q, \theta')$. As $\theta' \geq 2/(q+1)$, we obtain that

$$\begin{aligned} 316 \quad & \hat{F}_0\left(\frac{2}{q+1}\right) = L\left(\frac{2}{q+1}\right) = 2 - \frac{2(q-1)^2 \log(q-1)}{(q-2)q(q+1)}, \\ 317 \quad & G\left(\frac{2}{q+1}\right) = \log\left(\frac{q+1}{2}\right). \end{aligned}$$

It is enough to determine the sign of

$$\begin{aligned} 319 \quad & (\ln q) \left(\hat{F}_0\left(\frac{2}{q+1}\right) - G\left(\frac{2}{q+1}\right) \right) \\ 320 \quad &= \ln\left(\frac{2q}{q+1}\right) - \frac{2(q-1)^2}{(q-2)q(q+1)} \ln(q-1) =: \Delta(q). \end{aligned}$$

Compute directly $\Delta(3) = \ln\frac{3}{2} - \frac{2}{3} \ln 2 < 0$ and $\Delta(4) = \ln\frac{8}{5} - \frac{9}{20} \ln 3 < 0$, thus the maximizer of M is the solution

of the equation $L(\theta) = G(\theta)$ in I . In the cases $q \geq 5$, we estimate

$$\begin{aligned}\Delta(q) &= \ln 2 + \ln \left(1 - \frac{1}{q+1}\right) \\ &\quad - \left(1 + \frac{1}{q(q-2)}\right) \frac{2 \ln(q-1)}{q} \\ &\geq \ln 2 + \ln \left(1 - \frac{1}{6}\right) - \left(1 + \frac{1}{15}\right) \frac{2 \ln 4}{5} > 0,\end{aligned}$$

thus the maximizer of M is simply $\frac{2}{q+1}$ and the maximum of M is $G(2/(q+1))$. \square

Remark 2: Observe that $R(\mathcal{E}_f) \leq \frac{1}{2} \log \binom{q+1}{2}$ is a naive bound since the output alphabet of the two-user union channel has cardinality $\binom{q+1}{2}$. In [2, Sec. II], it was stated that equality holds in this naive bound for $q \geq 6$. It was then conjectured there that in Theorem 1 a random variable U of cardinality q readily gives $(R, R) \in \mathcal{E}_f$. However, the random variable U used in our proof of Theorem 4 has cardinality $2q$, and we do not see a way to reduce its cardinality.

III. ZERO-ERROR CAPACITY WITH COMPLETE FEEDBACK

In this section, we describe a zero-error communication scheme for the two-user union channel with complete feedback and show that it achieves a near-optimal rate pair $(R, R) \in \mathcal{O}_f$. This scheme, like that in [14, Sec. IV], partitions the uses of the channel into a large number $B + 1$ of blocks, each of length n except the last block. Suppose the message sets of the senders are both $[q]^{Bm}$, where $m \in [n]$ will be decided later, and let $w_1, w_2 \in [q]^{Bm}$ be the messages of the two senders.

To describe the communication scheme in each block, we represent the uncertainty of the receiver about the first bm digits of w_1 and w_2 at the end of block b by $U(b) \subseteq ([q] \times [q])^{bm}$. In other words, the receiver, at the end of block b , knows that

$$((w_{1i}, w_{2i}))_{i=1}^{bm} \in U(b).$$

The key idea of our communication scheme is to keep the uncertainty sets uniformly bounded in size.

Due to the feedback of the channel, the uncertainty set is common knowledge between the senders and the receiver. The initial uncertainty set $U(0) := \emptyset$. In addition, we assume for a moment that

at the end of block b each sender knows

the first bm digits of the other message. (10)

This assumption will be shown below to hold by induction on the blocks.

A. Indexing

At the start of block $b + 1$, the senders and the receiver index the elements in $U(b)$ by

$$S := \{(s_1, \dots, s_n) : s_k \in \{\ast\} \cup [q] \text{ such that } |\{k : s_k = \ast\}| = m\}.$$

We shall choose $m \in [n]$ carefully in Theorem 5 so that $|U(b)| \leq |S| = \binom{n}{m} q^{n-m}$. The method of indexing can be agreed beforehand between the senders and the receiver. For example, they can order both $U(b)$ and S lexicographically, and index the elements in the ordered set $U(b)$ by the first $|U(b)|$ elements in S .

B. Encoding

During block $b + 1$, according to the inductive assumption (10) of our scheme, both senders know $((w_{1i}, w_{2i}))_{i=1}^{bm}$ and its index $(s_1, \dots, s_n) \in S$. During the k th use of the channel in block $b + 1$, both senders simply send s_k if $s_k \in [q]$; and send $w_{1,bm+i}$ and $w_{2,bm+i}$ respectively if s_k is the i th star in (s_1, \dots, s_n) . In the latter case, based on the feedback, each sender learns the $(bm+i)$ th digit of the other message. Because there are a total of m stars in (s_1, \dots, s_n) , at the end of block $b + 1$, each sender learns m more digits of the other message, maintaining the inductive assumption (10) of the scheme.

C. Decoding

After n uses of the channel in block $b + 1$, the receiver has received $(y_1, \dots, y_n) \in (\mathcal{Y}_1 \cup \mathcal{Y}_2)^n$, where $\mathcal{Y}_i := \{y \subset [q] : |y| = i\}$. The receiver then enumerates (s_1, \dots, s_n) through the first $|U(b)|$ elements in S , and for each (s_1, \dots, s_n) compatible with the output, namely $y_k = \{s_k\}$ if $s_k \in [q]$, the receiver adds to $U(b + 1)$ all the $((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{(b+1)m} \in ([q] \times [q])^{(b+1)m}$ such that

- 1) $((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{bm} \in U(b)$ is indexed by (s_1, \dots, s_n) ; and
- 2) $\{\hat{w}_{1,bm+i}, \hat{w}_{2,bm+i}\} = y_{k_i}$ for all $i \in [m]$, where k_1, \dots, k_m are the indices of s_k that are stars.

The updated uncertainty set $U(b + 1)$ will be shown in Theorem 5 to be bounded by $|S|$ in size.

After B blocks of uses of the channel, the receiver obtains the uncertainty set $U(B)$, and both senders know (w_1, w_2) and its index $(s_1, \dots, s_n) \in S$ as a member of $U(B)$. Finally, in the last block $B + 1$, the senders simply communicate the index (s_1, \dots, s_n) through $\lceil \log |S| \rceil \leq n - m + \lceil \log \binom{n}{m} \rceil$ uses of the channel.

Theorem 5: If $n/2 \leq m \leq n$ and

$$\binom{2n-2m}{n-m} 2^{2m-n} \leq \binom{n}{m} q^{n-m}, \quad (11)$$

then the communication scheme described above allows the receiver to recover the messages $w_1, w_2 \in [q]^{Bm}$ from the senders without errors through $\leq Bn + n - m + \lceil \log \binom{n}{m} \rceil$ uses of the two-user union channel with complete feedback. In particular, $R(\mathcal{O}_f) \geq R$, where R is the solution of $H_b(\alpha) + (1-\alpha) \log_2 q = 1$ in $(1/2, 1]$, and H_b is the binary entropy function.

Proof: It suffices to show that $|U(b)| \leq |S|$ for all $b = 0, 1, \dots, B$. The base case is evident as $U(0)$ consists of the empty sequence. For the inductive step, assume that $|U(b)| \leq |S|$. During the $(b + 1)$ st block, the receiver has received $(y_1, \dots, y_n) \in (\mathcal{Y}_1 \cup \mathcal{Y}_2)^n$.

We shall estimate the size of the uncertainty set $U(b + 1)$ at the end of the $(b + 1)$ st block. Suppose that $((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{bm} \in U(b)$ is indexed by (s_1, \dots, s_n) . Recall that if $s_k \in$

[q], then $y_k = \{s_k\}$; otherwise s_k is the i th star and $\{\hat{w}_{1,bm+i}, \hat{w}_{2,bm+i}\} = y_k$. In other words, only when $y_k = \{s_k\}$ for every k such that $s_k \in [q]$, the uncertainty set $U(b+1)$ would include the following $\prod_{i=1}^m |y_{k_i}|$ elements

$\{((\hat{w}_{1i}, \hat{w}_{2i}))_{i=1}^{(b+1)m}\}$ such that

$$\{\hat{w}_{1,bm+i}, \hat{w}_{2,bm+i}\} = y_{k_i} \text{ for all } i \in [m],$$

where k_1, \dots, k_m are the indices of s_k that are stars. Suppose $L := \{k : y_k \in \mathcal{Y}_2\}$ and $\ell := |L|$. A $(s_1, \dots, s_n) \in S$ compatible with (y_1, \dots, y_n) must have stars on coordinates indexed by L and choose from the rest $n - \ell$ positions an additional $m - \ell$ stars. This (s_1, \dots, s_n) , if it indexes an element in $U(b)$, will contribute at most 2^ℓ elements to $U(b+1)$. Therefore, we can estimate $|U(b+1)| \leq \binom{n-\ell}{m-\ell} 2^\ell$.

We claim that this estimate $u_\ell := \binom{n-\ell}{m-\ell} 2^\ell$ reaches its maximum $\binom{2n-2m}{n-m} 2^{2m-n}$ at $\ell = 2m-n, 2m-n+1$. In fact, we compare

$$\frac{u_\ell}{u_{\ell+1}} = \frac{\binom{n-\ell}{m-\ell} 2^\ell}{\binom{n-\ell-1}{m-\ell-1} 2^{\ell+1}} = \frac{n-\ell}{2(m-\ell)},$$

which is less than 1 when $\ell < 2m-n$ and greater than 1 when $\ell > 2m-n$. Combining with (11), it is guaranteed that $|U(b+1)| \leq |S|$. This finishes the inductive step.

Finally we prove the lower bound on $R(\mathcal{O}_f)$. Fix $\alpha \in (1/2, 1]$ such that $H_b(\alpha) + (1-\alpha) \log_2 q > 1$. It is well known that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log_2 \binom{n}{\lceil \alpha n \rceil} = H_b(\alpha).$$

We can choose n sufficiently large and $m = \lceil \alpha n \rceil$ so that

$$\begin{aligned} \frac{1}{n} \log_2 \binom{n}{m} + \left(1 - \frac{m}{n}\right) \log_2 q &\geq 1 \\ \implies \binom{n}{m} q^{n-m} &\geq 2^n \geq \binom{2n-2m}{n-m} 2^{2m-n}. \end{aligned}$$

Our communication scheme provides a $(q^{Bm}, q^{Bm}, Bn+n-m + \lceil \log \binom{n}{m} \rceil, 0)$ code. Thus

$$R(\mathcal{O}_f) \geq \frac{Bm}{Bn+n-m+\lceil \log \binom{n}{m} \rceil} \rightarrow \frac{m}{n} \geq \alpha \text{ as } B \rightarrow \infty.$$

Note that $H_b(\alpha) + (1-\alpha) \log_2 q - 1$ is decreasing on $(1/2, 1]$ and has a unique root $R \in [0, 1]$. We can choose α arbitrarily close to the root R to show that $R(\mathcal{O}_f) \geq R$. \square

Remark 3: For $q = 2$, our communication scheme could theoretically achieve $R(\mathcal{O}_f) \geq 0.77291$. In practice, in order to achieve $R(\mathcal{O}_f) \geq 0.764$, we can choose $m = 13, n = 17, B = 1019$ for our scheme to obtain a $(2^{13247}, 2^{13247}, 17339, 0)$ code. During the encoding and decoding process, the senders and the receiver need to keep track of up to 35840 binary numbers of length 13247 in the uncertainty set, which takes up 59.35 megabytes of memory for storage.

Remark 4: For $q \geq 5$, a naive upper bound on $R(\mathcal{O}_f)$ is $R(\mathcal{E}_f) = \frac{1}{2} \log \binom{q+1}{2} = 1 - \frac{1}{2 \log_2 q} + O(1/q)$. Notice that when $\alpha = 1 - \frac{1}{\log_2 q}$, $H_b(\alpha) + (1-\alpha) \log_2 q > (1-\alpha) \log_2 q = 1$. The proof of Theorem 5 implies that $R(\mathcal{O}_f) \geq 1 - \frac{1}{\log_2 q}$. Then

gap between the upper bound and the lower bound on $R(\mathcal{O}_f)$ is about $\frac{1}{2 \log_2 q}$.

IV. OPEN PROBLEMS

At the moment, the channel capacity $R(\mathcal{E}_f)$ of the union channel with complete feedback is determined for every $q \geq 2$. The zero-error capacity $R(\mathcal{O}_f)$ however is yet to be determined for any q . Naturally, the first step for future research is to determine $R(\mathcal{O}_f)$ for $q = 2$. Based on the characterization of \mathcal{O}_f by Dueck [20], our numerical experiments suggest that that for the binary case the lower bound on $R(\mathcal{O}_f)$ proved by Belokopytov [21] is tight.

Conjecture A: The average zero-error capacity of the two-user binary adder channel, that is the union channel with $q = 2$, with complete feedback is equal to 0.78974.

An inspection of Table I reveals that $R(\mathcal{E}) < R(\mathcal{O}_f)$ for $q \leq 6$. However, for $q \geq 14$, the lower bound on $R(\mathcal{O}_f)$ in Theorem 5 is less than $R(\mathcal{E})$. We speculate that our lower bound on $R(\mathcal{O}_f)$ can be improved for every $q \geq 2$.

Conjecture B: For all $q \geq 2$, the average channel capacity $R(\mathcal{E})$ of the two-user union channel without feedback is strictly less than the average zero-error capacity $R(\mathcal{O}_f)$ of the two-user union channel with complete feedback.

Conjecture B would establish the chain of inequalities $R(\mathcal{O}) \leq R(\mathcal{E}) < R(\mathcal{O}_f) \leq R(\mathcal{E}_f)$.

APPENDIX

Proof of Lemma 2: Let X_1 and X_2 be independent $[q]$ -valued random variables so that $\Pr(X_1 = X_2) = \theta$, and let $a_i := \Pr(X_1 = i)$ and $b_i := \Pr(X_2 = i)$ for every $i \in [q]$. Then $F(\theta)$ is

$$\text{the maximum of } H(a_1, \dots, a_q) + H(b_1, \dots, b_q), \quad (12a)$$

$$\text{subject to } \sum a_i = 1, \sum b_i = 1, \sum a_i b_i = \theta, \quad a_i, b_i \geq 0. \quad (12b)$$

Clearly $F(\theta)$ is continuous with respect to θ .

We first prove the unimodality of F , that is, for any $\theta \in [0, 1]$, if θ' is between θ and $1/q$, then $F(\theta') \geq F(\theta)$. Let $\mathbf{a} = (a_1, \dots, a_q), \mathbf{b} = (b_1, \dots, b_q)$ be a maximizer of the optimization problem (12). Let $\mathbf{j} = (1/q, \dots, 1/q) \in \mathbb{R}^q$, and consider $\mathbf{a}' = (1-t)\mathbf{a} + t\mathbf{j}$ and $\mathbf{b}' = (1-t)\mathbf{b} + t\mathbf{j}$, where $t \in [0, 1]$ is a solution of

$$\begin{aligned} \theta' &= \mathbf{a}' \cdot \mathbf{b}' = ((1-t)\mathbf{a} + t\mathbf{j}) \cdot ((1-t)\mathbf{b} + t\mathbf{j}) \\ &= (1-t)^2 \theta + \frac{(2-t)t}{q}. \end{aligned} \quad (13)$$

The right hand side of (13) equals θ and $1/q$ when $t = 0, 1$ respectively. By the intermediate value theorem, (13) has a solution in $[0, 1]$ and t is well-defined. Note that \mathbf{a}', \mathbf{b}' satisfy the constraint (12b). By the concavity of the entropy function, we have

$$\begin{aligned} F(\theta') &\geq H(\mathbf{a}') + H(\mathbf{b}') \geq (1-t)H(\mathbf{a}) + tH(\mathbf{j}) \\ &\quad + (1-t)H(\mathbf{b}) + tH(\mathbf{j}) \geq H(\mathbf{a}) + H(\mathbf{b}) = F(\theta). \end{aligned}$$

From this point forward, we assume that $\theta \in (1/q, 1]$ is fixed. We shall repeatedly add constraints to the optimization problem (12) without decreasing its maximum.

Given \mathbf{a}, \mathbf{b} satisfying the constraint (12b), consider the vectors \mathbf{a}', \mathbf{b}' whose coordinates are respectively the ones of \mathbf{a}, \mathbf{b} sorted in non-decreasing order. The rearrangement inequality says

$$\theta' := \left(\frac{\mathbf{a}' + \mathbf{b}'}{2} \right) \cdot \left(\frac{\mathbf{a}' + \mathbf{b}'}{2} \right) \geq \mathbf{a}' \cdot \mathbf{b}' \geq \mathbf{a} \cdot \mathbf{b} = \theta.$$

As θ is between $1/q$ and θ' , again we have $\mathbf{a}'' = \mathbf{b}'' = (1-t)\frac{\mathbf{a}'+\mathbf{b}'}{2} + t\mathbf{j}$ such that $\mathbf{a}'' \cdot \mathbf{b}'' = \theta$ for some $t \in [0, 1]$. By the concavity of the entropy function, we have

$$\begin{aligned} H(\mathbf{a}'') + H(\mathbf{b}'') &\geq 2 \left((1-t)H\left(\frac{\mathbf{a}'+\mathbf{b}'}{2}\right) + tH(\mathbf{j}) \right) \\ &\geq 2H\left(\frac{\mathbf{a}'+\mathbf{b}'}{2}\right) \geq H(\mathbf{a}') + H(\mathbf{b}') \\ &= H(\mathbf{a}) + H(\mathbf{b}). \end{aligned}$$

We come to the conclusion that without loss of generality we may assume $\mathbf{a} = \mathbf{b}$ in (12). The optimization problem is then equivalent to

$$\text{Maximize: } \frac{2}{\ln q} \left(- \sum_{i=1}^q x_i \ln x_i \right), \quad (14a)$$

$$\text{Subject to: } \sum_{i=1}^q x_i = 1, \quad \sum_{i=1}^q x_i^2 = \theta, \quad x_i \geq 0 \quad \text{for all } i \in [q]. \quad (14b)$$

Consider the Lagrangian

$$\begin{aligned} \mathcal{L}(\mathbf{x}, \lambda_1, \lambda_2) &:= - \sum_{i=1}^q x_i \ln x_i \\ &\quad + \lambda_1 \left(\sum_{i=1}^q x_i - 1 \right) + \lambda_2 \left(\sum_{i=1}^q x_i^2 - \theta \right). \end{aligned}$$

The method of Lagrange multipliers gives necessary conditions for the maximizers:

$$\frac{\partial \mathcal{L}}{\partial x_i} = -\ln x_i - 1 + \lambda_1 + 2\lambda_2 x_i = 0, \quad \text{for all } i \in [q].$$

In other words, given λ_1, λ_2 , each coordinate of a maximizer \mathbf{x} is a solution of the equation $-\ln x_i - 1 + \lambda_1 + 2\lambda_2 x_i = 0$. Since $x \mapsto -\ln x - 1 + \lambda_1 + 2\lambda_2 x$ is convex, there are at most two solutions. Without loss of generality, we can add to (14b) the constraints that

$$x_1 = \dots = x_r = a, \quad x_{r+1} = \dots = x_q = b,$$

for some $r \in [q-1]$, and $a \geq b \geq 0$. We have thus reduced the optimization problem (14a) to

$$\text{Maximize: } \frac{2}{\ln q} (-ra \ln a - sb \ln b), \quad (15a)$$

$$\text{Subject to: } ra + sb = 1, \quad ra^2 + sb^2 = \theta, \quad (15b)$$

$$r + s = q, \quad r, s \in [q], \quad a \geq b \geq 0. \quad (15c)$$

Given r, s such that $r + s = q$, we can solve from $ra + sb = 1, ra^2 + sb^2 = \theta$ for a, b :

$$\begin{aligned} (a, b) &= \left(\frac{1}{q} + \frac{\sqrt{rsp}}{qr}, \frac{1}{q} - \frac{\sqrt{rsp}}{qs} \right) \\ &\quad \text{or } \left(\frac{1}{q} - \frac{\sqrt{rsp}}{qr}, \frac{1}{q} + \frac{\sqrt{rsp}}{qs} \right). \end{aligned}$$

where $p := \theta q - 1 \in (0, q-1]$. Because $a \geq b$, we discard the second solution of (a, b) . Given the parameter $t := r/q \in [1/q, 1-1/q]$, the variables $r = qt, s = q - qt$ and

$$a = \frac{1}{q} \left(1 + \sqrt{\frac{1-t}{t} p} \right), \quad b = \frac{1}{q} \left(1 - \sqrt{\frac{t}{1-t} p} \right) \quad (560)$$

can be seen as functions of t , so can the objective function $v(t) := -ra \ln a - sb \ln b$. The derivative of v is

$$v' = -q(a \ln a + ta'(\ln a + 1) - b \ln b + (1-t)b'(\ln b + 1)). \quad (563)$$

The implicit differentiation of $ra + sb = 1, ra^2 + sb^2 = \theta$ yields

$$\begin{aligned} q(a + ta' - b + (1-t)b') &= 0, \\ a^2 + 2taa' - b^2 + 2(1-t)bb' &= 0. \end{aligned} \quad (566) \quad (567)$$

which can be viewed as a system of linear equations of a', b' . Using $a > 1/q > b$, we deduce that

$$a' = -\frac{a-b}{2t}, \quad b' = -\frac{a-b}{2(1-t)}. \quad (570)$$

and we simplify v' as follows:

$$\begin{aligned} v' &= -q(a \ln a - \frac{1}{2}(a-b)(\ln a + 1) \\ &\quad - b \ln b - \frac{1}{2}(a-b)(\ln b + 1)) \\ &= -q\left(\frac{a+b}{2} \ln\left(\frac{a}{b}\right) + b - a\right) \\ &= qb\left(-\frac{1}{2}(r+1) \ln r - 1 + r\right), \end{aligned} \quad (572) \quad (573) \quad (574) \quad (575)$$

where $r = a/b > 1$. One can check that $-\frac{1}{2}(r+1) \ln r - 1 + r < 0$ for $r > 1$, and so $v(t)$ is decreasing. In other words, the objective function v attains its maximum at $t = 1/q$ or equivalently the maximizer of (15) is given by $r = 1, s = q-1, a = \alpha, b = \frac{1-\alpha}{q-1}$, which leads to $F(\theta) = 2H(\alpha, 1-\alpha; 1, q-1)$. \square

ACKNOWLEDGEMENTS

We thank the referees for the suggestions which improved both the exposition of the paper and the clarity of the proofs.

REFERENCES

- [1] S. C. Chang and J. K. Wolf, "On the T -user M -frequency noiseless multiple-access channel with and without intensity information," *IEEE Trans. Inf. Theory*, vol. IT-27, no. 1, pp. 41–48, Jan. 1981, doi: 10.1109/TIT.1981.1056304.
- [2] A. Vinck, W. Hoeks, and K. A. Post, "On the capacity of the two-user M -ary multiple-access channel with feedback (Corresp.)," *IEEE Trans. Inf. Theory*, vol. IT-31, no. 4, pp. 540–543, Jul. 1985, doi: 10.1109/TIT.1985.1057066.
- [3] T. M. Cover and J. A. Thomas, *Elements of Information Theory*, 2nd ed. Hoboken, NJ, USA: Wiley, 2006.
- [4] R. Ahlswede, "Combination methods models," in *Rudolf Ahlswede's Lectures on Information Theory* (Foundations in Signal Processing, Communications and Networking), vol. 13. A. Ahlswede, I. Althöfer, C. Deppe, and U. Tamm. Eds. Cham, Switzerland: Springer, 2018, doi: 10.1007/978-3-319-53139-7.
- [5] R. Ahlswede, "Multi-way communication channels," in *Proc. 2nd Int. Symp. Inf. Theory*. Tsahkadsor, Armenia, 1973, pp. 23–51.
- [6] H. H.-J. Liao, "Multiple access channels," Ph.D. dissertation, Dept. Elect. Eng., Univ. Hawaii, Honolulu, HI, USA, Sep. 1972.

- 605 [7] M. Mattas and P. R. J. Ostergard, "A new bound for the zero-
606 error capacity region of the two-user binary adder channel," *IEEE
607 Trans. Inf. Theory*, vol. 51, no. 9, pp. 3289–3291, Sep. 2005,
608 doi: 10.1109/TIT.2005.853309.
- 609 [8] B. Lindström, "Determination of two vectors from the sum," *J. Combinat. Theory*, vol. 6, no. 4, pp. 402–407, May 1969.
- 610 [9] G. Cohen, S. Litsyn, and G. Zémor, "Binary B_2 -sequences: A new
611 upper bound," *J. Combinat. Theory, Ser. A*, vol. 94, no. 1, pp. 152–155,
612 Apr. 2001. [Online]. Available: <http://www.sciencedirect.com/science/article/pii/S0097316500931273>
- 613 [10] F. Gao and G. Ge, "New bounds on separable codes for multimedia
614 fingerprinting," *IEEE Trans. Inf. Theory*, vol. 60, no. 9, pp. 5257–5262,
615 Sep. 2014, doi: 10.1109/TIT.2014.2331989.
- 616 [11] S. R. Blackburn, "Probabilistic existence results for separable codes,"
617 *IEEE Trans. Inf. Theory*, vol. 61, no. 11, pp. 5822–5827, Nov. 2015,
618 doi: 10.1109/TIT.2015.2473848 Nov.
- 619 [12] C. Shangguan, X. Wang, G. Ge, and Y. Miao, "New bounds for frame-
620 proof codes," *IEEE Trans. Inf. Theory*, vol. 63, no. 11, pp. 7247–7252,
621 Nov. 2017, doi: 10.1109/TIT.2017.2745619.
- 622 [13] N. Gaarder and J. K. Wolf, "The capacity region of a multiple-access
623 discrete memoryless channel can increase with feedback (Corresp.),"
624 *IEEE Trans. Inf. Theory*, vol. IT-21, pp. 100–102, 1975.
- 625 [14] T. Cover and C. S. K. Leung, "An achievable rate region for the multiple-
626 access channel with feedback," *IEEE Trans. Inf. Theory*, vol. IT-27,
627 no. 3, pp. 292–298, May 1981, doi: 10.1109/TIT.1981.1056357.
- 628 [15] F. M. J. Willems, "The feedback capacity region of a class of discrete
629 memoryless multiple access channels," *IEEE Trans. Inf. Theory*,
630 vol. IT-28, no. 1, pp. 93–95, Jun. 1982, doi: 10.1109/TIT.1982.1056437.
- 631 [16] M. Salehi, "Cardinality bounds on auxiliary variables in multiple-
632 user theory via the method of Ahlswede and Korner," Dept. Statist.,
633 Stanford Univ., Stanford, CA, USA, Tech. Rep. 33, 1978.
- 634 [17] F. Willems, "On multiple access channels with feedback (Corresp.),"
635 *IEEE Trans. Inf. Theory*, vol. IT-30, no. 6, pp. 842–845, Nov. 1984.
- 636 [18] V. V. Prelov, "On one extreme value problem for entropy and error
637 probability," *Problems Inf. Transmiss.*, vol. 50, no. 3, pp. 203–216,
638 Jun. 2014, doi: 10.1134/S003294601403016.
- 639 [19] V. V. Prelov, "On some extremal problems for mutual information and
640 entropy," *Problemy Peredachi Informatsii*, vol. 52, no. 4, pp. 3–13, 2016
641 doi: 10.1134/s0032946016040013.
- 642 [20] G. Dueck, "The zero error feedback capacity region of a certain class of
643 multiple-access channels," *Problems Control Inf. Theory*, vol. 14, no. 2,
644 pp. 89–103, 1985.
- 645 [21] A. Belokopytov, "On the zero error feedback capacity region of the
646 binary adder channel," *Problems Control Inf. Theory*, vol. 18, no. 2,
647 pp. 125–133, 1989.
- 648 [22] C. Christen, "A Fibonacci algorithm for the detection of two elements,"
649 Ph.D. dissertation, Dept. d'IRO, Univ. Montréal, Montréal, QC,
650 Canada, 1980
- 651 [23] M. Aigner, "Search problems on graphs," *Discrete Appl. Math.*, vol. 14,
652 no. 3, pp. 215–230, Jul. 1986, doi: 10.1016/0166-218X(86)90026-0.
653
- 654 [24] Z. Zhang, T. Berger, and J. Massey, "Some families of zero-error block
655 codes for the two-user binary adder channel with feedback," *IEEE Trans.
656 Inf. Theory*, vol. 33, no. 5, pp. 613–619, Sep. 1987.
657
- 658 [25] L. Gargano, V. Montour, G. Setaro, and U. Vaccaro, "An improved
659 algorithm for quantitative group testing," *Discrete Appl. Math.*,
660 vol. 36, no. 3, pp. 299–306, May 1992. [Online]. Available:
661 <http://www.sciencedirect.com/science/article/pii/0166218X9290260H>
662
- 663 [26] A. Y. Belokopytov and V. N. Luzgin, "Block transmission of information
664 in a summing multiple access channel with feedback," *Problems Inf.
665 Transmiss.*, vol. 23, no. 4, pp. 347–351, Apr. 1988.
666
- 667 [27] H. G. Eggleston, *Convexity* (Cambridge Tracts in Mathematics and
668 Mathematical Physics), vol. 47. New York, NY, USA: Cambridge Univ.
669 Press, 1958.
670
- 671
- 672
- 673
- Zilin Jiang is an applied mathematics instructor at Massachusetts Institute of Technology. Up until July 2018, he was a postdoctoral fellow at Technion–Israel Institute of Technology. He graduated from Peking University in 2011 with a B.Sc. degree in Mathematics, and received his Ph.D. in Algorithm, Combinatorics and Optimization from Carnegie Mellon University in 2016.
- Nikita Polyanskii was born in Russia in 1991. He received the M.Sc. degree in Mathematics and the Ph.D. degree in Mathematics from the Lomonosov Moscow State University, Moscow, Russia, in 2013 and 2016, respectively. During 2015–2017 he was a researcher at the Institute for Information Transmission Problems, Moscow, Russia, and a senior engineer at Huawei Technologies, Moscow, Russia. Since 2017 Nikita has been a postdoctoral researcher in the Department of Mathematics, Technion–Israel Institute of Technology, Haifa, Israel. Since 2018 he has been a research scientist in the Center for Computational and Data-Intensive Science and Engineering, Skolkovo Institute of Science and Technology, Moscow, Russia. His research interests include coding theory and its applications to communications, storage systems, and combinatorics.
- Ilya Vorobyev received the M.Sc. degree in Mathematics and the Ph.D. degree in Mathematics from the Lomonosov Moscow State University in 2013 and 2017, respectively. In 2015–2017 he worked as a research engineer at Huawei R&D department in Moscow. He also was a researcher at the Institute for Information Transmission Problems, Moscow, in 2015–2017. Since 2017 Ilya has been a senior researcher in the Advanced Combinatorics and Complex Networks Lab, Moscow Institute of Physics and Technology. Since 2018 he has been a research scientist in the Center for Computational and Data-Intensive Science and Engineering, Skolkovo Institute of Science and Technology. His research interests include extremal combinatorics and coding theory.