# Upper bound on list-decoding radius of binary codes

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Abstract—Consider the problem of packing Hamming balls of a given relative radius subject to the constraint that they cover any point of the ambient Hamming space with multiplicity at most L. For odd  $L \geq 3$  an asymptotic upper bound on the rate of any such packing is proven. The resulting bound improves the best known bound (due to Blinovsky'1986) for rates below a certain threshold. The method is a superposition of the linearprogramming idea of Ashikhmin, Barg and Litsyn (that was used previously to improve the estimates of Blinovsky for L = 2) and a Ramsey-theoretic technique of Blinovsky. As an application it is shown that for all odd L the slope of the rate-radius tradeoff is zero at zero rate.

*Index Terms*—Combinatorial coding theory, list-decoding, converse bounds

# I. MAIN RESULT AND DISCUSSION

One of the most well-studied problems in information theory asks to find the maximal rate at which codewords can be packed in binary space with a given minimum distance between codewords. Operationally, this (still unknown) rate gives the capacity of the binary input-output channel subject to adversarial noise of a given level. A natural generalization was considered by Elias and Wozencraft [1], [2], who allowed the decoder to output a list of size L. In this paper we provide improved upper bounds on the latter question.

Our interest in bounding the asymptotic tradeoff for the listdecoding problem is motivated by our study of fundamental limits of joint source-channel communication [3]. Namely, in [4, Theorem 6] we proposed an extension of the previous result in [3, Theorem 7] that required bounding rate for the list-decoding problem.

We proceed to formal definitions and brief overview of known results. For a binary code  $\mathcal{C} \subset \mathbb{F}_2^n$  we define its list-size L decoding radius as

$$\tau_L(\mathcal{C}) \stackrel{\triangle}{=} \frac{1}{n} \max\{r : \forall x \in \mathbb{F}_2^n \ |\mathcal{C} \cap \{x + B_r^n\}| \le L\},\$$

where Hamming ball  $B_r^n$  and Hamming sphere  $S_r^n$  are defined as

$$B_r^n \stackrel{\triangle}{=} \left\{ x \in \mathbb{F}_2^n : |x| \le r \right\},\tag{1}$$

$$S_r^n \stackrel{\triangle}{=} \{ x \in \mathbb{F}_2^n : |x| = r \}$$

$$\tag{2}$$

with  $|x| = |\{i : x_i = 1\}|$  denoting the Hamming weight of x. Alternatively, we may define  $\tau_L$  as follows:<sup>1</sup>

$$\tau_L(\mathcal{C}) = \frac{1}{n} \left( \min \left\{ \operatorname{rad}(S) : S \in \begin{pmatrix} \mathcal{C} \\ L+1 \end{pmatrix} \right\} - 1 \right),$$

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 $\binom{\mathcal{C}}{i}$  denotes the set of all subsets of  $\mathcal{C}$  of size j.

where rad(S) denotes radius of the smallest ball containing S (known as Chebyshev radius):

$$\operatorname{rad}(S) \stackrel{\triangle}{=} \min_{y \in \mathbb{F}_2^n} \max_{x \in S} |y - x|.$$

The asymptotic tradeoff between rate and list-decoding radius  $\tau_L$  is defined as usual:

$$\tau_L^*(R) \stackrel{\triangle}{=} \limsup_{n \to \infty} \max_{\mathcal{C}: |\mathcal{C}| \ge 2^{nR}} \tau_L(\mathcal{C}) \tag{3}$$

$$R_L^*(\tau) \stackrel{\triangle}{=} \limsup_{n \to \infty} \sup_{\mathcal{C}: \tau_L(\mathcal{C}) \ge \tau} \frac{1}{n} \log |\mathcal{C}| \tag{4}$$

The best known upper (converse) bounds on this tradeoff are as follows:

• List size L = 1: The best bound to date was found by McEliece, Rodemich, Rumsey and Welch [5]:

$$R_1^*(\tau) \le R_{LP2}(2\tau) \,, \tag{5}$$

$$R_{LP2}(\delta) \stackrel{\triangle}{=} \min \log 2 - h(\alpha) + h(\beta), \qquad (6)$$

where  $h(x) = -x \log x - (1-x) \log(1-x)$  and minimum is taken over all  $0 \le \beta \le \alpha \le 1/2$  satisfying

$$2\frac{\alpha(1-\alpha)-\beta(1-\beta)}{1+2\sqrt{\beta(1-\beta)}} \le \delta$$

For rates R < 0.305 this bound coincides with the simpler bound:

$$\tau_1^*(R) \le \frac{1}{2} \delta_{LP1}(R),$$
(7)

$$\delta_{LP1}(R) \stackrel{\triangle}{=} \frac{1}{2} - \sqrt{\beta(1-\beta)}, \quad R = \log 2 - h(\beta), \quad (8)$$

where  $\beta \in [0, \frac{1}{2}]$ .

• List size L = 2: The bound found by Ashikhmin, Barg and Litsyn [6] is given as<sup>2</sup>

$$R_2^*(\tau) \le \log 2 - h(2\tau) + R_{up}(2\tau, 2\tau),$$

where  $R_{up}(\delta, \alpha)$  is the best known upper bound on rate of codes with minimal distance  $\delta n$  constrained to live on Hamming spheres  $S_{\alpha n}^n$ . The expression for  $R_{up}(\delta, \alpha)$ can be obtained by using the linear programming bound from [5] and applying Levenshtein's monotonicity, cf. [7, Lemma 4.2(6)]. The resulting expression is

$$R_{2}^{*}(\tau) \leq \begin{cases} R_{LP2}(2\tau), & \tau \leq \tau_{0} \\ \log 2 - h(2\tau) + h(u(\tau)), & \tau > \tau_{0}, \end{cases}$$
(9)

where  $\tau_0 \approx 0.1093$  and

$$u(\tau) = \frac{1}{2} - \sqrt{\frac{1}{4} - (\sqrt{\tau - 3\tau^2} - \tau)^2}$$

<sup>2</sup>This result follows from optimizing [6, Theorem 4]. It is slightly stronger than what is given in [6, Corollary 5].

(cf. [7, (9)]).

For list sizes L ≥ 3: The original bound of Blinovsky [8] appears to be the best (before this work):

$$\tau_L^*(R) \le \sum_{i=1}^{\lceil L/2 \rceil} \frac{\binom{2i-2}{i-1}}{i} (\lambda(1-\lambda))^i, \qquad R = 1 - h(\lambda),$$
(10)

where  $\lambda \in [0, \frac{1}{2}]$ . Note that [8] also gives a nonconstructive lower bound on  $\tau_L^*(R)$ . Results on listdecoding over non-binary alphabets are also known, see [9], [10].

In this paper we improve the bound of Blinovsky for lists of odd size and rates below a certain threshold. To that end we will mix the ideas of Ashikhmin, Barg and Litsyn (namely, extraction of a large spectrum component from the code) and those of Blinovsky (namely, a Ramsey-theoretic reduction to study of symmetric subcodes).

To present our main result, we need to define exponent of Krawtchouk polynomial  $K_{\beta n}(\xi n) = \exp\{nE_{\beta}(\xi) + o(n)\}$ . For  $\xi \in [0, \frac{1}{2} - \sqrt{\beta(1-\beta)}]$  the value of  $E_{\beta}(\xi)$  was found in [11]. Here we give it in the following parametric form, cf. [12] or [13, Lemma 4]:

$$E_{\beta}(\xi) = \xi \log(1-\omega) + (1-\xi)\log(1+\omega) - \beta \log\omega \quad (11)$$

$$\xi = \frac{1}{2} (1 - (1 - \beta)\omega - \beta\omega^{-1}), \qquad (12)$$

where

$$\omega \in \left[\frac{\beta}{1-\beta}, \sqrt{\frac{\beta}{1-\beta}}\right]$$

Our main result is the following:

**Theorem 1.** Fix list size  $L \ge 2$ , rate R and an arbitrary  $\beta \in [0, 1/2]$  with  $h(\beta) \le R$ . Then any sequence of codes  $C_n \subset \{0, 1\}^n$  of rate R satisfies

$$\limsup_{n \to \infty} \tau_L(\mathcal{C}_n) \le \\ \max_{j,\xi_0} \xi_0 g_j \left( 1 - \frac{\xi_1}{2\xi_0} \right) + (1 - \xi_0) g_j \left( \frac{\xi_1}{2(1 - \xi_0)} \right) , \quad (13)$$

where maximization is over  $\xi_0$  satisfying

$$0 \le \xi_0 \le \frac{1}{2} - \sqrt{\beta(1-\beta)}$$
 (14)

and j ranging over  $\{0, 1, 3, ..., 2k + 1, ..., L\}$  if L is odd and over  $\{0, 2, ..., 2k, ..., L\}$  if L is even. Quantity  $\xi_1 = \xi_1(\xi_0, \delta, R)$  is a unique solution of

$$R + h(\beta) - 2E_{\beta}(\xi_{0}) = h(\xi_{0}) - \xi_{0}h\left(\frac{\xi_{1}}{2\xi_{0}}\right) - (1 - \xi_{0})h\left(\frac{\xi_{1}}{2(1 - \xi_{0})}\right), \quad (15)$$

on the interval  $[0, 2\xi_0(1-\xi_0)]$  and functions  $g_j(\nu)$  are defined as

$$g_j(\nu) \stackrel{\triangle}{=} \frac{1}{L+j} \left( L\nu - \mathbb{E}\left[ |2W - L - j|^+ \right] \right), W \sim \operatorname{Bino}(L,\nu)$$
(16)

As usual with bounds of this type, cf. [14], it appears that taking  $h(\beta) = R$  can be done without loss. Under such choice,

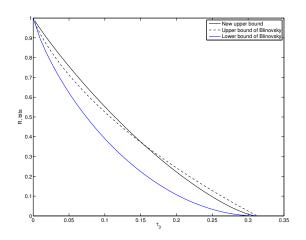


Fig. 1. Comparison of bounds on  $R_L^*(\tau)$  for list size L = 3

TABLE I RATES FOR WHICH NEW BOUND\* IMPROVES STATE OF THE ART

List size $L$	Range of rates
L = 3	$0 < R \le 0.361$
L = 5	$0 < R \le 0.248$
L = 7	$0 < R \le 0.184$
L = 9	$0 < R \le 0.136$
L = 11	$\begin{array}{c} 0 < R \leq 0.361 \\ 0 < R \leq 0.248 \\ 0 < R \leq 0.184 \\ 0 < R \leq 0.136 \\ 0 < R \leq 0.100 \end{array}$

\* This is computation of (13) with  $h(\beta) = R$ .

our bound outperforms Blinovsky's for all odd L and all rates small enough (see Corollary 3 below). The bound for L = 3 is compared in Fig. 1 with the result of Blinovsky numerically. For larger odd L the comparison is similar, but the range of rates where our bound outperforms Blinovsky's becomes smaller, see Table I.

Evaluation of Theorem 1 is computationally possible, but is somewhat tedious. Fortunately, for small *L* the maximum over  $\xi_0$  and *j* is attained at  $\xi_0 = \frac{1}{2} - \sqrt{\beta(1-\beta)}$  and j = 1. We rigorously prove this for L = 3.<sup>3</sup>

**Corollary 2.** For list-size L = 3 we have

$$\tau_L^*(R) \le \frac{3}{4}\delta - \frac{1}{16} \left( \frac{(2\delta - \xi_1)^3}{\delta^2} + \frac{\xi_1^3}{(1 - \delta)^2} \right), \quad (17)$$

where  $\delta \in (0, 1/2]$  and  $\xi_1 \in [0, 2\delta(1 - \delta)]$  are functions of R determined from

$$R = h\left(\frac{1}{2} - \sqrt{\delta(1-\delta)}\right), \qquad (18)$$

$$R = \log 2 - \delta h\left(\frac{\xi_1}{2\delta}\right) - (1-\delta)h\left(\frac{\xi_1}{2(1-\delta)}\right)$$
(19)

Another interesting implication of Theorem 1 is that it allows us to settle the question of slope of the curve  $R_L^*(\tau)$  at zero rate. Notice that Blinovsky's converse bound (10) has a negative slope, while his achievability bound has a zero slope. Our bound always has a zero slope for odd L (but not for even L, see Remark 2 in Section II-C):

<sup>3</sup>Notice that proofs of each of the two Corollaries below contain different relaxations of the bound (13), e.g. (22), which are easier to evaluate. Notice also that in Table I for the last two entries (L = 9, 11) at the high endpoint of rate the maximum over  $\xi_0$  is attained *not* at  $\frac{1}{2} - \sqrt{\beta(1-\beta)}$ .

**Corollary 3.** Fix arbitrary odd  $L \ge 3$ . There exists  $R_0 = R_0(L) > 0$  such that for all rates  $R < R_0$  we have

$$\tau_L^*(R) \le g_1(\delta_{LP1}(R)),$$
 (20)

where  $g_1(\cdot)$  is a degree-L polynomial defined in (16). In particular,

$$\frac{d}{d\tau}\Big|_{\tau=\tau_L^*(0)} R_L^*(\tau) = 0\,, \tag{21}$$

where the zero-rate radius is  $\tau_L^*(0) = \frac{1}{2} - 2^{-L-1} \left(\frac{L}{L-1}\right)$ .

Before closing our discussion we make some additional remarks:

- The bound in Theorem 1 can be slightly improved by replacing δ<sub>LP1</sub>(R), that appears in the right-hand side of (14), with a better bound, a so-called second linear-programming bound δ<sub>LP2</sub>(R) from [5]. This would enforce the usage of the more advanced estimate of Litsyn [15, Theorem 5] and complicate analysis significantly. Notice that δ<sub>LP2</sub>(R) ≠ δ<sub>LP1</sub>(R) only for rates R ≥ 0.305. If we focus attention only on rates where new bound is better than Blinovsky's, such a strengthening only affects the case of L = 3 and results in a rather minuscule improvement (for example, for rate R = 0.33 the improvement is ≈ 3 · 10<sup>-5</sup>).
- 2) For even L it appears that  $h(\beta) = R$  is no longer optimal. However, the resulting bound does not appear to improve upon Blinovsky's.
- 3) When L is large (e.g. 35) the maximum in (13) is not always attained by either j = 1 or  $\xi_0 = \delta_{LP1}(R)$ . It is not clear whether such anomalies only happen in the region of rates where our bound is inferior to Blinovsky's.
- The result of Corollary 3 follows by weakening (13) (via concavity of g<sub>j</sub>, Lemma 8) to

$$\limsup_{n \to \infty} \tau_L(\mathcal{C}_n) \le \max_{j,\xi_0} g_j(\xi_0) = \max_j g_j(\delta_{LP1}(R)).$$
(22)

The  $R < R_0(L)$  condition is only used to show that the maximum is attained at j = 1. Note also that weakening (22) corresponds to omitting the extra Elias-Bassalygo type reduction, which is responsible for the extra optimization over  $\xi_1$  in (13).

Finally, at the invitation of anonymous reviewer we give our intuition about why our bound outperforms Blinovsky's for odd L. It is easiest to compare with the weakening (22) of our bound. Now compare the two proofs:

- Blinovsky [8] first uses Elias-Bassalygo reduction to restrict attention to a subcode C' situated on a Hamming sphere of radius ≈ δ<sub>GV</sub>(R) = h<sup>-1</sup>(1 − R). Then he proves an upper bound for τ<sub>L</sub>(C') valid as long as |C'| ≫ 1 via a Plotkin-type argument together with a great symmetrization idea.
- Our bound (following Ashikhmin, Barg and Litsyn [6]) instead uses a Kalai-Linial [11] reduction to select a subcode C" situated on a Hamming sphere of radius

 $\approx \delta_{LP1}(R)$ . We then proceeded to prove a (Plotkintype) upper bound on a strange quantity:

$$\tau_L^o(\mathcal{C}'') = \frac{1}{n} \left( \min\left\{ \operatorname{rad}(\{0\} \cup S) : S \in \binom{\mathcal{C}}{L} \right\} - 1 \right) \,,$$

which corresponds to a requirement that the code contain not more than L-1 codewords in any ball of radius  $\tau_L^o$ , but only for those balls that happen to also contain the origin.

Notice that the sphere returned by Kalai-Linial is bigger than that of Elias-Bassalygo (which is the reason our bound deteriorates at large rates), but the good thing is that the subcode C'' has another codeword  $c_0$  at the center of the Hamming sphere. Now, intuitively  $\tau_L^o$  is roughly equivalent to  $\tau_{L-1}$ . The zero-rate (Plotkin) radius for a list-L decoding of binary codes on Hamming sphere  $S^n_{\varepsilon_n}$  is given by

$$p_L(\xi) = \frac{\mathbb{E}\left[\min(W_{\xi}, L+1-W_{\xi})\right]}{L+1}, W_{\xi} \sim \operatorname{Bino}(L+1, \xi).$$

So intuitively, we expect that Blinovsky's bound should give

 $\tau_L^*(R) \lesssim p_L(\delta_{GV}(R))$ 

while our bound should give

$$\tau_L^*(R) \lesssim p_{L-1}(\delta_{LP1}(R))$$
.

Finally, it is easy to check that for even L we have  $p_L = p_{L-1}$ , while for odd L,  $p_L > p_{L-1}$ . This is the main intuitive reason why our bound succeeds in improving Blinovsky's, but only for odd L.

#### II. PROOFS

# A. Proof of Theorem 1

Consider an arbitrary sequence of codes  $C_n$  of rate R. As in [6] we start by using Delsarte's linear programming to select a large component of the distance distribution of the code. Namely, we apply result of Kalai and Linial [11, Proposition 3.2]: For every  $\beta$  with  $h(\beta) \leq R$  there exists a sequence  $\epsilon_n \rightarrow 0$  such that for every code C of rate R there is a  $\xi_0$ satisfying (14) such that

$$A_{\xi_0 n}(\mathcal{C}) \stackrel{\triangle}{=} \frac{1}{|\mathcal{C}|} \sum_{x, x' \in \mathcal{C}} 1\{|x - x'| = \xi_0 n\}$$
$$\geq \exp\{n(R + h(\beta) - 2E_\beta(\xi_0) + \epsilon_n)\}. \quad (23)$$

Without loss of generality (by compactness of the interval  $[0, 1/2 - \sqrt{\beta(1-\beta)}]$  and passing to a proper subsequence of codes  $C_{n_k}$ ) we may assume that  $\xi_0$  selected in (23) is the same for all blocklengths n. Then there is a sequence of subcodes  $C'_n$  of asymptotic rate

$$R' \ge R + h(\beta) - 2E_{\beta}(\xi_0)$$

such that each  $C'_n$  is situated on a sphere  $c_0 + S_{\xi_0}$  surrounding another codeword  $c_0 \in C$ . Our key geometric result is: If there are too many codewords on a sphere  $c_0 + S_{\xi_0}$  then it is possible to find L of them that are includable in a small ball that also contains  $c_0$ . Precisely, we have: **Lemma 4.** Fix  $\xi_0 \in (0, 1)$  and positive integer L. There exist a sequence  $\epsilon_n \to 0$  such that for any code  $C'_n \subset S_{\xi_0 n}$  of rate R' > 0 there exist L codewords  $c_1, \ldots, c_L \in C'_n$  such that

$$\frac{1}{n} \operatorname{rad}(0, c_1, \dots, c_L) \le \theta(\xi_0, R', L) + \epsilon_n , \qquad (24)$$

where

$$\theta(\xi_0, R', L) \stackrel{\triangle}{=} \max_j \theta_j(\xi_0, R', L) \tag{25}$$

$$\theta_j(\xi_0, R', L) \stackrel{\triangle}{=} \xi_0 g_j \left( 1 - \frac{\xi_1}{2\xi_0} \right) + (1 - \xi_0) g_j \left( \frac{\xi_1}{2(1 - \xi_0)} \right),$$
(26)

with  $\xi_1 = \xi_1(\xi_0)$  found as unique solution on interval  $[0, 2\xi_0(1 - \xi_0)]$  of

$$R' = h(\xi_0) - \xi_0 h\left(\frac{\xi_1}{2\xi_0}\right) - (1 - \xi_0) h\left(\frac{\xi_1}{2(1 - \xi_0)}\right), \quad (27)$$

functions  $g_j$  are defined in (16) and j in maximization (25) ranging over the same set as in Theorem 1.

Equipped with Lemma 4 we immediately conclude that

$$\limsup_{n \to \infty} \tau_L(\mathcal{C}_n) \le \max_{\xi_0 \in [0,\delta]} \theta(\xi_0, R + h(\beta) - 2E_\beta(\xi_0), L) .$$
(28)

Clearly, (28) coincides with (13). So it suffices to prove Lemma 4.

#### B. Proof of Lemma 4

Let  $\mathcal{T}_L$  be the  $(2^L - 1)$ -dimensional space of probability distributions on  $\mathbb{F}_2^L$ . If  $T \in \mathcal{T}_L$  then we have

$$T = (t_v, v \in \mathbb{F}_2^L) \qquad t_v \ge 0, \sum_v t_v = 1$$

We define distance on  $\mathcal{T}_L$  to be the  $L_\infty$  one:

$$||T - T'|| \stackrel{\triangle}{=} \max_{v \in \mathbb{F}_2^L} |t_v - t'_v|$$

Permutation group  $S_L$  acts naturally on  $\mathbb{F}_2^L$  and this action descends to probability distributions  $\mathcal{T}_L$ . We will say that T is symmetric if

$$T = \sigma(T) \quad \iff \quad t_v = t_{\sigma(v)} \quad \forall v \in \mathbb{F}_2^L$$

for any permutation  $\sigma : [L] \to [L]$ . Note that symmetric T is completely specified by L + 1 numbers (weights of Hamming spheres in  $\mathbb{F}_2^L$ ):

$$\sum_{v:|v|=j} t_v, \qquad j=0,\ldots,L.$$

Next, fix some total ordering of  $\mathbb{F}_2^n$  (for example, lexicographic). Given a subset  $S \subset \mathbb{F}_2^n$  we will say that S is given in ordered form if  $S = \{x_1, \ldots, x_{|S|}\}$  and  $x_1 < x_2 \cdots < x_{|S|}$ under the fixed ordering on  $\mathbb{F}_2^n$ . For any subset of codewords  $S = \{x_1, \ldots, x_L\}$  given in ordered form we define its *joint* type T(S) as an element of  $\mathcal{T}_L$  with

$$t_v \stackrel{\triangle}{=} \frac{1}{n} |\{j : x_1(j) = v_1, \dots, x_L(j) = v_j\}|,$$

where here and below y(j) denotes the *j*-th coordinate of binary vector  $y \in \mathbb{F}_2^n$ . In this way every subset S is associated

to an element of  $\mathcal{T}_L$ . Note that T(S) is symmetric if and only if the  $L \times n$  binary matrix representing S (by combining row-vectors  $x_j$ ) has the property that the number of columns equal to  $[1, 0, ..., 0]^T$  is the same as the number of columns  $[0, 1, ..., 0]^T$  etc. For any code  $\mathcal{C} \subset \mathbb{F}_2^n$  we define its average joint type:

$$\bar{T}_L(\mathcal{C}) = \frac{1}{L! \cdot \binom{|\mathcal{C}|}{L}} \sum_{\sigma} \sum_{S \in \binom{\mathcal{C}}{L}} \sigma(T(S)) \,.$$

Evidently,  $\overline{T}_L(\mathcal{C})$  is symmetric.

Our proof crucially depends on a (slight extension of the) brilliant idea of Blinovsky [8]:

**Lemma 5.** For every  $L \ge 1$ ,  $K \ge L$  and  $\delta > 0$  there exist a constant  $K_1 = K_1(L, K, \delta)$  such that for all  $n \ge 1$  and all codes  $\mathcal{C} \subset \mathbb{F}_2^n$  of size  $|\mathcal{C}| \ge K_1$  there exists a subcode  $\mathcal{C}' \subset \mathcal{C}$ of size at least K such that for any  $S \in \binom{\mathcal{C}'}{L}$  we have

$$\|T(S) - \overline{T}_L(\mathcal{C}')\| \le \delta.$$
<sup>(29)</sup>

**Remark 1.** Note that if  $S' \subset S$  then every element of T(S') is a sum of  $\leq 2^L$  elements of T(S). Hence, joint types T(S') are approximately symmetric also for smaller subsets |S'| < L.

*Proof.* We first will show that for any  $\delta_1 > 0$  and sufficiently large  $|\mathcal{C}|$  we may select a subcode  $\mathcal{C}'$  so that the following holds: For any pair of subsets  $S, S' \subset \mathcal{C}'$  s.t.  $|S| = |S'| \leq L$  we have:

$$||T(S) - T(S')|| \le \delta_1 \tag{30}$$

Consider any code  $C_1 \subset \mathbb{F}_2^n$  and define a hypergraph with vertices indexed by elements of C and hyper-edges corresponding to each of the subsets of size L. Now define a  $\delta_1/2$ -net on the space  $\mathcal{T}_L$  and label each edge according to the closest element of the  $\delta_1/2$ -net. By a theorem of Ramsey there exists  $K_L$  such that if  $|C_1| \geq K_L$  then there is a subset  $C'_1 \subset C$  such that  $|C'_1| \geq K$  and each of the internal edges, indexed by  $\binom{C'_1}{L}$ , is assigned the same label. Thus, by triangle inequality (30) follows for all  $S, S' \in \binom{C'_1}{L}$ .

Next, apply the previous argument to show that there is a constant  $K_{L-1}$  such that for any  $C_2 \subset \mathbb{F}_2^n$  of size  $|C_2| \ge K_{L-1}$  there exists a subcode  $C'_2$  of size  $|C'_2| \ge K_L$  satisfying (30) for all  $S, S' \in \binom{C'_2}{L-1}$ . Since  $C'_2$  satisfies the size assumption on  $C_1$  made in previous paragraph, we can select a further subcode  $C''_2 \subset C'_2$  of size  $\ge K_L$  so that for  $C''_2$  property (30) holds for all S, S' of size L or L-1.

Continuing similarly, we may select a subcode C' of arbitrary C such that (30) holds for all  $|S| = |S'| \leq L$  provided that  $|C| \geq K_1$ .

Next, we show that (30) implies

$$||T(S_0) - \sigma(T(S_0))|| \le C\delta_1$$
, (31)

where  $S_0 \in \binom{\mathcal{C}'}{L}$  is arbitrary and C = C(L) is a constant depending on L only.

Now to prove (31) let  $T(S_0) = \{t_v, v \in \mathbb{F}_2^L\}$  and consider an arbitrary transposition  $\sigma : [L] \to [L]$ . It will be clear that our proof does not depend on what transposition is chosen, so for simplicity we take  $\sigma = \{(L-1) \leftrightarrow L\}$ . We want to show that (30) implies

$$|t_v - t_{\sigma(v)}| \le \delta_1 \,. \qquad \forall v \in \mathbb{F}_2^L \tag{32}$$

Since transpositions generate permutation group  $S_L$ , (31) then follows. Notice that (32) is only informative for v whose last two digits are not equal, say  $v = [v_0, 0, 1]$ . Suppose that  $S_0 = \{c_1, \ldots, c_L\}$  given in the ordered form. Let

$$S = \{c_1, \dots c_{L-1}\},$$
 (33)

$$S' = \{c_1, \dots, c_{L-2}, c_L\}$$
(34)

Joint types T(S) and T(S') are expressible as functions of  $T(S_0)$  in particular, the number of occurrences of element  $[v_0, 0]$  in S is  $t_{[v_0, 0, 1]} + t_{[v_0, 0, 0]}$  and in S' is  $t_{[v_0, 0, 0]} + t_{[v_0, 1, 0]}$ . Thus, from (30) we obtain:

$$|(t_{[v_0,0,1]} + t_{[v_0,0,0]}) - (t_{[v_0,0,0]} + t_{[v_0,1,0]})| \le \delta$$

implying (32) and thus (31).

Finally, we show that (31) implies (29). Indeed, consider the chain

$$\|T(S) - \bar{T}_L(\mathcal{C}')\|$$

$$= \left\|T(S) - \frac{1}{L! \cdot \binom{|\mathcal{C}'|}{L}} \sum_{\sigma} \sum_{S' \in \binom{\mathcal{C}'}{L}} \sigma(T(S'))\right\| \quad (35)$$

$$\leq \frac{1}{L! \cdot \binom{|\mathcal{C}'|}{L}} \sum_{\sigma} \sum_{S' \in \binom{\mathcal{C}'}{L}} \|T(S) - \sigma(T(S'))\| \qquad (36)$$

$$\leq \frac{1}{L! \cdot \binom{|\mathcal{C}'|}{L}} \sum_{\sigma} \sum_{S' \in \binom{\mathcal{C}'}{L}} \|T(S) - T(S')\| \\ + \|T(S') - \sigma(T(S'))\|$$

$$\leq (1+C)\delta_1, \qquad (38)$$

where (36) is by convexity of the norm, (37) is by triangle inequality and (38) is by (30) and (31). Consequently, setting  $\delta_1 = \frac{\delta}{1+C}$  we have shown (29).

Before proceeding further we need to define the concept of an average radius (or a moment of inertia):

$$\overline{\mathrm{rad}}(x_1,\ldots,x_m) \stackrel{\triangle}{=} \min_y \frac{1}{m} \sum_{i=1}^m |x_i - y|.$$

Note that the minimizing y can be computed via a percoordinate majority vote (with arbitrary tie-breaking for even m). Consider now an arbitrary subset  $S = \{c_1, \ldots, c_L\}$  and define for each  $j \ge 0$  the following functions

$$h_j(S) \stackrel{\triangle}{=} \frac{1}{n} \overline{\mathrm{rad}}(\underbrace{0, \dots, 0}_{j \text{ times}}, c_1, \dots, c_L).$$

It is easy to find an expression for  $h_j(S)$  in terms of the jointtype of S:

$$h_j(S) = \frac{1}{L+j} \left( \mathbb{E}[W] - \mathbb{E}[|2W - L - j|^+] \right)$$
(39)

$$\mathbb{P}[W=w] = \sum_{v:|v|=w} t_v , \qquad (40)$$

where  $t_v$  are components of the joint-type  $T(S) = \{t_v, v \in \mathbb{F}_2^L\}$ . To check (39) simply observe that if one arranges L codewords of S in an  $L \times n$  matrix and also adds j rows of zeros, then computation of  $h_j(S)$  can be done per-column: each column of weight w contributes

$$\min(w, L + j - w) = w - |2w - L - j|^+$$

to the sum. In view of expression (39) we will abuse notation and write

$$h_j(T(S)) \stackrel{\triangle}{=} h_j(S)$$

We now observe that for symmetric codes satisfying (29) average-radii  $h_j(S)$  in fact determine the regular radius:

**Lemma 6.** Consider an arbitrary code C satisfying conclusion (29) of Lemma 5. Then for any subset  $S = \{c_1, \ldots, c_L\} \subset C$  we have

$$\left| \operatorname{rad}(0, c_1, \dots, c_L) - n \cdot \max_j h_j(\bar{T}_L(\mathcal{C})) \right| \le 2^L (1 + \delta n),$$
(41)

where j in maximization (41) ranges over  $\{0, 1, 3, \ldots, 2k + 1, \ldots, L\}$  if L is odd and over  $\{0, 2, \ldots, 2k, \ldots, L\}$  if L is even.

*Proof.* For joint-types of size L and all  $j \ge 0$  we clearly have (cf. expression (39))

$$|h_j(T_1) - h_j(T_2)| \le 2^{L-1} ||T_1 - T_2||, \qquad \forall T_1, T_2 \in \mathcal{T}_L.$$
(42)

We also trivially have

$$\frac{1}{n}\operatorname{rad}(0, c_1, \dots, c_L) \ge h_j(S) \qquad \forall j \ge 0.$$
(43)

Thus from (29) and (42) we already get

$$\frac{1}{n} \operatorname{rad}(0, c_1, \dots, c_L) \ge \max_j h_j(\bar{T}_L(\mathcal{C})) - 2^{L-1}\delta.$$

It remains to show

$$\frac{1}{n}\operatorname{rad}(0,c_1,\ldots,c_L) \le \max_j h_j(\bar{T}_L(\mathcal{C})) + \delta + \frac{2^L}{n}.$$
 (44)

This evidently requires constructing a good center y for the set  $\{0, c_1, \ldots, c_L\}$ . To that end fix arbitrary numbers  $q = (q_0, \ldots, q_L) \in [0, 1]^L$ . Next, for each  $v \in \mathbb{F}_2^L$  let  $E_v \subset [n]$ be all coordinates on which restriction of  $\{c_1, \ldots, c_L\}$  equals v. On  $E_v$  put y to have a fraction  $q_{|v|}$  of ones and remaining set to zeros (rounding to integers arbitrarily). Proceed for all  $v \in \mathbb{F}_2^L$ . Call resulting vector  $y(q) \in \mathbb{F}_2^n$ .

Denote for convenience  $c_0 = 0$ . We clearly have

$$\operatorname{rad}(c_0, c_1, \dots, c_L) \le \min_{q} \max_{p} \sum_{i=0}^{L} p_i |c_i - y(q)|, \quad (45)$$

where  $p = (p_0, \dots, p_L)$  is a probability distribution. Denote

$$T(S) = \{t_v, v \in \mathbb{F}_2^L\}$$

$$\tag{46}$$

$$\bar{T}_L(\mathcal{C}) = \{ \bar{t}_v, v \in \mathbb{F}_2^L \}$$
(47)

We proceed to computing  $|c_i - y(q)|$ .

$$|c_{i} - y(q)| \leq n \sum_{v \in \mathbb{F}_{2}^{L}} t_{v}(q_{|v|} 1\{v(i) = 0\} + (1 - q_{|v|}) 1\{v(i) = 1\}) + 2^{L}, \quad (48)$$

where  $2^L$  comes upper-bounding the integer rounding issues and we abuse notation slightly by setting v(0) = 0 for all v(recall that v(i) is the *i*-th coordinate of  $v \in \mathbb{F}_2^L$ ).

By (29) we may replace  $t_v$  with  $\bar{t}_v$  at the expense of introducing  $2^L \delta n$  error, so we have:

$$\begin{aligned} |c_i - y(q)| &\leq n \sum_{v \in \mathbb{F}_2^L} \bar{t}_v(q_{|v|} 1\{v(i) = 0\} \\ &+ (1 - q_{|v|}) 1\{v(i) = 1\}) + 2^L (1 + \delta n) \,. \end{aligned}$$
(49)

Next notice that the sum over v only depends on whether i = 0 or  $i \neq 0$  (by symmetry of  $\bar{t}_v$ ). Furthermore, for any given weight w and  $i \neq 0$  we have

$$\sum_{v:|v|=w} 1\{v(i)=1\} = \binom{L}{w} \frac{w}{L}.$$

Thus, introducing the random variable  $\overline{W}$ , cf. (39),

$$\mathbb{P}[\bar{W} = w] \stackrel{\triangle}{=} \sum_{v:|v|=w} \bar{t}_v ,$$

we can rewrite:

$$\sum_{v \in \mathbb{F}_2^L} \bar{t}_v(q_{|v|} 1\{v(i) = 0\} + (1 - q_{|v|}) 1\{v(i) = 1\})$$
$$= \frac{1}{L} \mathbb{E}\left[\bar{W} + (L - 2\bar{W})q_{\bar{W}}\right]. \quad (50)$$

For i = 0 the expression is even simpler:

$$\sum_{v \in \mathbb{F}_2^L} \bar{t}_v(q_{|v|} 1\{v(0) = 0\} + (1 - q_{|v|}) 1\{v(0) = 1\}) = \mathbb{E}\left[q_{\bar{W}}\right].$$

Substituting derived upper bound on  $|c_i - y(q)|$  into (45) we can see that without loss of generality we may assume  $p_1 = \cdots = p_L$ , so our upper bound (modulo  $O(\delta)$  terms) becomes:

$$\min_{q} \max_{p_1 \in [0, L^{-1}]} (1 - Lp_1) \mathbb{E}[q_{\bar{W}}] + p_1 \mathbb{E}[\bar{W} + (L - 2\bar{W})q_{\bar{W}}]$$
  
= 
$$\min_{q} \max_{p_1 \in [0, L^{-1}]} p_1 \mathbb{E}[\bar{W}] + \mathbb{E}[q_{\bar{W}}(1 - 2\bar{W}p_1)]$$

By von Neumann's minimax theorem we may interchange min and max, thus continuing as follows:

$$= \max_{p_1 \in [0, L^{-1}]} \min_{q} p_1 \mathbb{E}\left[\bar{W}\right] + \mathbb{E}\left[q_{\bar{W}}(1 - 2\bar{W}p_1)\right]$$
(51)

$$= \max_{p_1 \in [0,L^{-1}]} p_1 \mathbb{E}\left[\bar{W}\right] - \mathbb{E}\left[|2\bar{W}p_1 - 1|^+\right].$$
 (52)

The optimized function of  $p_1$  is piecewise-linear, so optimization can be reduced to comparing values at slopediscontinuities and boundaries. The point  $p_1 = 0$  is easily excluded, while the rest of the points are given by  $p_1 = \frac{1}{L+i}$  with j ranging over the set specified in the statement of Lemma<sup>4</sup>. So we continue (52) getting

$$= \max_{j} \frac{1}{L+j} \left( \mathbb{E}\left[\bar{W}\right] - \mathbb{E}\left[|2\bar{W} - L - j|^{+}\right] \right)$$
(53)

We can see that expression under maximization is exactly  $h_j(\bar{T}_L(\mathcal{C}))$  and hence (44) is proved.

**Lemma 7.** There exist constants  $C_1, C_2$  depending only on L such that for any  $C \subset \mathbb{F}_2^n$  the joint-type  $\overline{T}_L(C)$  is approximately a mixture of product Bernoulli distributions<sup>5</sup>, namely:

$$\left\|\bar{T}_L(\mathcal{C}) - \frac{1}{n} \sum_{i=1}^n \operatorname{Bern}^{\otimes L}(\lambda_i)\right\| \le \frac{C_1}{|\mathcal{C}|}, \quad (54)$$

where  $\lambda_i = \frac{1}{|\mathcal{C}|} \sum_{c \in \mathcal{C}} 1\{c(i) = 1\}$  be the density of ones in the *j*-th column of a  $|\mathcal{C}| \times n$  matrix representing the code. In particular,

$$\left| h_j(\bar{T}_L(\mathcal{C})) - \frac{1}{n} \sum_j g_j(\lambda_j) \right| \le \frac{C_2}{|\mathcal{C}|},$$
 (55)

where functions  $g_j$  were defined in (16).

*Proof.* Second statement (55) follows from the first via (42) and linearity of  $h_j(T)$  in the type T, cf. (39). To show the first statement, let  $M = |\mathcal{C}|$ ,  $M_i = \lambda_i M$  and  $p_w$  – total probability assigned to vectors v of weight w by  $\overline{T}_L(\mathcal{C})$ . Then by computing  $p_w$  over columns of  $M \times n$  matrix we obtain

$$p_w = \frac{1}{n} \sum_{i=1}^n \frac{\binom{M_i}{w} \binom{M-M_i}{L-w}}{\binom{M}{L}}.$$

By a standard estimate we have for all  $w = \{0, ..., L\}$ :

$$\frac{\binom{M_i}{w}\binom{M-M_i}{L-w}}{\binom{M}{L}} = \binom{L}{w}\lambda_i^w(1-\lambda_i)^{L-w} + O(\frac{1}{M}),$$

with  $O(\cdot)$  term uniform in w and  $\lambda_i$ . By symmetry of the type  $\overline{T}_L(\mathcal{C})$  the result (54) follows.

**Lemma 8.** Functions  $g_j$  defined in (16) are concave on [0, 1].

*Proof.* Let  $W_{\lambda} \sim \text{Bino}(L, \lambda)$  and  $V_{\lambda} \sim \text{Bino}(L - 1, \lambda)$ . Denote for convenience  $\overline{\lambda} = 1 - \lambda$  and take  $j_0$  to be an integer

<sup>&</sup>lt;sup>4</sup>The difference between odd and even L occurs due to the boundary point  $p_1 = \frac{1}{L}$  not being a slope-discontinuity when L is odd, so we needed to add it separately.

<sup>&</sup>lt;sup>5</sup>Distribution Bern<sup> $\otimes L$ </sup>( $\lambda$ ) assigns probability  $\lambda^{|v|}(1-\lambda)^{L-|v|}$  to element  $v \in \mathbb{F}_2^L$ .

between 0 and L. We have then

$$\frac{\partial}{\partial\lambda} \mathbb{E}\left[|W_{\lambda} - j_{0}|^{+}\right]$$

$$= \sum_{w=j_{0}+1}^{L} {L \choose w} (w - j_{0}) \lambda^{w} \bar{\lambda}^{L-w} \left\{ w \lambda^{-1} - (L-w) \bar{\lambda}^{-1} \right\}$$
(56)

$$= {\binom{L}{j_0+1}} (j_0+1)\lambda^{j_0}\bar{\lambda}^{L-j_0-1} + \sum_{w=j_0+1}^{L-1} \left[ {\binom{L}{w+1}} (w+1-j_0)(w+1) - {\binom{L}{w}} (w-j_0)(L-w) \right] \lambda^w \bar{\lambda}^{L-w-1}$$
(57)

$$= L \binom{L-1}{j_0} \lambda^{j_0} \bar{\lambda}^{L-1-j_0} + L \sum_{w=j_0+1}^{L-1} \binom{L-1}{w} \lambda^w \bar{\lambda}^{L-1-w}$$
(58)

$$= L\mathbb{P}[V_{\lambda} \ge j_0], \qquad (59)$$

where in (57) we shifted the summation by one for the first term under the sum in (56), and in (58) applied identities  $\binom{L}{w+1} = \binom{L}{w} \frac{L-w}{w+1} = \binom{L-1}{w} \frac{L}{w+1}$ . Similarly, if  $\theta \in [0,1)$  we have

$$\frac{\partial}{\partial\lambda}\mathbb{E}\left[|W_{\lambda}-j_{0}-\theta|^{+}\right] = L\mathbb{P}[V_{\lambda} \ge j_{0}+1] + L(1-\theta)\mathbb{P}[V_{\lambda}=j_{0}].$$
(60)

Similarly, one shows (we will need it later in Lemma 9):

$$\frac{\partial}{\partial \lambda} \mathbb{P}[W_{\lambda} \ge j_0] = L \mathbb{P}[V_{\lambda} = j_0 - 1].$$
(61)

Since clearly the function in (60) is strictly increasing in  $\lambda$  for any  $j_0$  and  $\theta$  we conclude that

$$\lambda \mapsto \mathbb{E}\left[|W_{\lambda} - j_0 - \theta|^+\right]$$

is convex. This concludes the proof of concavity of  $g_j$ .

Proof of Lemma 4. Our plan is the following:

- Apply Elias-Bassalygo reduction to pass from C'<sub>n</sub> to a subcode C''<sub>n</sub> on an intersection of two spheres S<sub>ξ0n</sub> and y + S<sub>ξ1n</sub>.
- 2) Use Lemma 5 to pass to a symmetric subcode  $C_n'' \subset C_n''$
- Use Lemmas 7-8 to estimate maxima of average radii h<sub>j</sub> over C<sup>'''</sup><sub>n</sub>.
- 4) Use Lemma 6 to transport statement about  $h_j$  to a statement on  $\tau_L(\mathcal{C}_n^{\prime\prime\prime})$ .

We proceed to details. It is sufficient to show that for some constant C = C(L) and arbitrary  $\delta > 0$  estimate (24) holds with  $\epsilon_n = C\delta$  whenever  $n \ge n_0(\delta)$ . So we fix  $\delta > 0$  and consider a code  $\mathcal{C}' \subset S_{\xi_0 n} \subset \mathbb{F}_2^n$  with  $|\mathcal{C}'| \ge \exp\{nR' + o(n)\}$ . Note that for any r, even m with  $m/2 \le \min(r, n - r)$  and arbitrary  $y \in S_r^n$  intersection  $\{y + S_m^n\} \cap S_r^n$  is isometric to the product of two lower-dimensional spheres:

$$\{y + S_m^n\} \cap S_r^n \cong S_{r-m/2}^r \times S_{m/2}^{n-r}$$
. (62)

Therefore, we have for  $r = \xi_0 n$  and valid m:

$$\sum_{y \in S_r^n} |\{y + S_m^n\} \cap \mathcal{C}'| = |\mathcal{C}'| \binom{\xi_0 n}{\xi_0 n - m/2} \binom{n(1 - \xi_0)}{m/2}.$$

Consequently, we can select  $m = \xi_1 n - o(n)$ , where  $\xi_1$  defined in (27), so that for some  $y \in S_r^n$ :

$$|\{y + S_{on}^n\} \cap \mathcal{C}'| > n.$$

Note that we focus on solution of (27) satisfying  $\xi_1 < 2\xi_0(1 - \xi_0)$ . For some choices of  $R, \delta$  and  $\xi_0$  choosing  $\xi_1 > 2\xi_0(1 - \xi_0)$  is also possible, but such a choice appears to result in a weaker bound.

Next, we let  $C'' = \{y + S_{\rho n}^n\} \cap C'$ . For sufficiently large n the code C'' will satisfy assumptions of Lemma 5 with  $K \ge \frac{1}{\delta}$ . Denote the resulting large symmetric subcode C'''.

Note that because of (62) column-densities  $\lambda_i$ 's of C''', defined in Lemma 7, satisfy (after possibly reordering coordinates):

$$\sum_{i=1}^{\xi_0 n} \lambda_i = \xi_1 n/2 + o(n), \quad \sum_{i > \xi_0 n} \lambda_i = \xi_1 n/2 + o(n).$$

Therefore, from Lemmas 7-8 we have

$$h_{j}(\bar{T}_{L}(\mathcal{C}''')) \leq \xi_{0}g_{j}\left(1 - \frac{\xi_{1}}{2\xi_{0}}\right) + (1 - \xi_{0})g_{j}\left(\frac{\xi_{1}}{2(1 - \xi_{0})}\right) + \epsilon'_{n} + \frac{C_{1}}{|\mathcal{C}'''|}, \quad (63)$$

where  $\epsilon'_n \to 0$ . Note that by construction the last term in (63) is  $O(\delta)$ . Also note that the first two terms in (63) equal  $\theta_j$  defined in (25).

Finally, by Lemma 6 we get that for any codewords  $c_1, \ldots, c_L \in \mathcal{C}'''$ , some constant C and some sequence  $\epsilon''_n \to 0$  the following holds:

$$\frac{1}{n} \operatorname{rad}(0, c_1, \dots, c_L) \le \theta(\xi_0, R', L) + \epsilon_n'' + C\delta.$$

By the initial remark, this concludes the proof of Lemma 4.  $\hfill \Box$ 

# C. Proof of Corollary 3

**Lemma 9.** For any odd L = 2a + 1 there exists a neighborhood of  $x = \frac{1}{2}$  such that

$$\max_{i} g_j(x) = g_1(x),$$
 (64)

maximum taken over j equal all the odd numbers not exceeding L and j = 0. We also have for some c > 0

$$g_1(x) = \frac{1}{2} - 2^{-L-1} \binom{L}{\frac{L-1}{2}} + cx + O((2x-1)^2), \qquad x \to \frac{1}{2}.$$
(65)

*Proof.* First, the value  $g_1(1/2)$  is computed trivially. Then from (60) we have

$$\frac{d}{dx}g_j(x) = \frac{L}{L+j}\left(1 - 2\mathbb{P}\left[V_x \ge \frac{L+j}{2}\right]\right),\tag{66}$$

where  $j \ge 1$  and  $V_x \sim \text{Bino}(x, L-1)$ . This implies (65). For future reference we note that (69) (below) and (61) imply

$$\frac{d}{dx}g_0(x) = 1 - 2\mathbb{P}[V_x \ge \frac{L+1}{2}] - \mathbb{P}[V_x = \frac{L-1}{2}],$$
$$V_x \sim \text{Bino}(x, L-1). \quad (67)$$

By continuity, (64) follows from showing

$$g_1(1/2) > \max_{j \in \{0,3,5,\dots L\}} g_j(1/2)$$
. (68)

Next, consider  $W_x \sim \text{Bino}(x, L)$  and notice the upper-bound

$$g_j(x) \le \frac{1}{L+j} \mathbb{E} \left[ W_x \mathbb{1}\{W_x \le a\} + (L+j-W_x) \mathbb{1}\{W_x \ge a\} \right]$$

Then, substituting expression for  $g_1(x)$  we get

$$g_1(x) - g_0(x) = \frac{1}{L} \left( \mathbb{P}[W_x \ge a+1] - g_1(x) \right)$$
(69)

$$g_1(x) - g_j(x) \ge \frac{j-1}{L+j} \left( g_1(x) - \mathbb{P}[W_x > a+1] \right)$$
. (70)

Thus, to show (68) it is sufficient to prove that for x = 1/2 we have

$$\mathbb{P}[W_{\frac{1}{2}} > a+1] < g_1(1/2) < \mathbb{P}[W_{\frac{1}{2}} \ge a+1].$$
 (71)

The right-hand inequality is trivial since  $\mathbb{P}[W_{\frac{1}{2}} \ge a+1] = 1/2$  while from (65) we know  $g_1(1/2) < 1/2$ . The left-hand inequality, after simple algebra, reduces to showing

$$\sum_{u=0}^{a-1} (2a+1-2u) \binom{2a+1}{u} < (2a+1) \binom{2a+1}{a}.$$
 (72)

Notice, that

$$(n-2u)\binom{n}{u} = n\left[\binom{n-1}{u} - \binom{n-1}{u-1}\right] \forall u \ge 0$$

and therefore

$$\sum_{u \le \ell} (n-2u) \binom{n}{u} = n \binom{n-1}{\ell}.$$

Plugging this identity into the right-hand side of (72) we get

$$\sum_{u=0}^{a-1} (2a+1-2u) \binom{2a+1}{u} = (2a+1) \binom{2a}{a-1} < (2a+1) \binom{2a}{a} < (2a+1) \binom{2a}{a} < (2a+1) \binom{2a+1}{a}$$
(73)

completing the proof of (72).

*Proof of Corollary 3.* We first show that (20) implies (21). To that end, fix a small  $\epsilon >$ so that  $\frac{1}{2} - \epsilon$  belongs to the neighborhood existence of which is claimed in Lemma 9. Choose rate so that  $\delta_{LP1}(R) = 1/2 - \epsilon$  and notice that this implies

$$R = h(\epsilon^2 + o(\epsilon^2)), \qquad (74)$$

By Lemma 9, the right-hand side of (20) is

$$\tau_L^*(0) - \operatorname{const} \cdot \epsilon + o(\epsilon),$$

which together with (74) implies (21).

To prove (20) we use Theorem 1 with  $\delta = \delta_{LP1}(R)$ . Next, use concavity of  $g_j$ 's (Lemma 8) to relax (13) to

$$\limsup_{n \to \infty} \tau_L(\mathcal{C}_n) \le \max_{j,\xi_0} g_j(\xi_0)$$

From (66) and (67) it is clear that  $\xi_0 \mapsto g_j(\xi_0)$  is monotonically increasing for all  $j \ge 0$  on the interval [0, 1/2]. Thus, we further have

$$\limsup_{n \to \infty} \tau_L(\mathcal{C}_n) \le \max_j g_j(\delta_{LP1}(R)).$$
(75)

+ Bdund (75) is valid for all  $R \in [0, 1]$  and arbitrary (odd/even L). However, when R is small (say,  $R < R_0$ ) and L is odd,  $\delta_{LP1}(R)$  belongs to the neighborhood of 1/2 in Lemma 9 and thus (20) follows from (75) and (64).

**Remark 2.** It is, perhaps, instructive to explain why Corollary 3 cannot be shown for even L (via Theorem 1). For even L the maximum over j of  $g_j(1/2 - \epsilon)$  is attained at j = 0 and

$$g_0(\frac{1}{2} - \epsilon) = \tau_L^*(0) + c\epsilon^2 + O(\epsilon^3), \epsilon \to 0$$
 (76)

Therefore, for  $\delta_{LP1}(R) = \frac{1}{2} - \epsilon$  we get from (76) that the right-hand side of (75) evaluates to

$$\tau_L^*(0) - \operatorname{const} \cdot \epsilon^2 \log \frac{1}{\epsilon}$$
 (77)

Thus, comparing (77) with (74) we conclude that for even L our bound on  $R_L^*(\tau)$  has negative slope at zero rate. Note that Blinovsky's bound (10) has negative slope at zero rate for both odd and even L.

# D. Proof of Corollary 2

*Proof.* Instead of working with parameter  $\delta$  we introduce  $\beta \in [0, 1/2]$  such that

$$\delta = \frac{1}{2} - \sqrt{\beta(1-\beta)} \,.$$

We then apply Theorem 1 with  $h(\beta) = R$ . Notice that the bound on  $\xi_0$  in (14) becomes

$$0 \le \xi_0 \le \delta$$

By a simple substitution  $\omega = \sqrt{\frac{\beta}{1-\beta}}$  we get from (11)

$$E_{\beta}(\delta) = \frac{1}{2} (\log 2 - h(\delta) + h(\beta)).$$

Therefore, when  $\xi_0 = \delta$  we notice that

$$R + h(\beta) - 2E_{\beta}(\xi_0) = R - \log 2 + h(\delta)$$

implying that defining equation for  $\xi_1$ , i.e. (15), coincides with (19).

Next for L = 3 we compute

$$g_0(\nu) = \nu(1-\nu),$$
 (78)

$$g_1(\nu) = \frac{3}{4}\nu - \frac{1}{2}\nu^3, \qquad (79)$$

$$g_3(\nu) = \frac{1}{2}\nu.$$
 (80)

Note that the right-hand side of (17) is precisely equal to

$$\delta g_1\left(1-\frac{\xi_1}{2\delta}\right)+(1-\delta)g_1\left(\frac{\xi_1}{2(1-\delta)}\right)$$

So this corollary simply states that for L = 3 the maximum in (13) is achieved at  $j = 1, \xi_0 = \delta$ . Let us restate this last statement rigorously: The maximum

$$\max_{j \in \{0,1,3\}} \max_{\xi_0 \in \delta} \xi_0 g_j \left( 1 - \frac{x}{2\xi_0} \right) + (1 - \xi_0) g_j \left( \frac{x}{2(1 - \xi_0)} \right)$$
(81)

is achieved at  $j = 1, \xi_0 = \delta$ . Here  $x = x(\xi_0, \beta)$  is a solution of

$$2(h(\beta) - E_{\beta}(\xi_0)) = h(\xi_0) - \xi_0 h\left(\frac{x}{2\xi_0}\right) - (1 - \xi_0) h\left(\frac{x}{2(1 - \xi_0)}\right).$$
(82)

For notational convenience we will denote the function under maximization in (81) by  $g_j(\xi_0, x)$ .

We proceed in two steps:

• First, we estimate the maximum over  $\xi_0$  for j = 0 as follows:

$$\max_{\xi_0} g_0(\xi_0, x) \le \frac{\log 2 - R}{4 \log 2} \cdot \left( 1 - \frac{1 - \delta}{a_{max}(1 - a_{max})} \right) + (1 - \delta)g_0(a_{min}), \quad (83)$$

where  $a_{max}, a_{min} \leq \frac{1}{2}$  are given by

$$a_{max} = h^{-1} (\log 2 - R), \qquad (84)$$

$$a_{min} = h^{-1} \left( \log 2 - \frac{R}{1 - \delta} \right)$$
 (85)

• Second, we prove that for j = 1 function

$$\xi_0 \mapsto g_j(\xi_0, x(\xi_0))$$

is monotonically increasing.

Once these two steps are shown, it is easy to verify (for example, numerically) that  $g_1(\delta, x(\delta))$  exceeds both  $\frac{1}{2}\delta$  (term corresponding to j = 3 in (81)) and the right-hand side of (83) (term corresponding to j = 0). Notice that this relation holds for all rates. Therefore, maximum in (81) is indeed attained at  $j = 1, \xi_0 = \delta$ .

One trick that will be common to both steps is the following. From the proof of Lemma 4 it is clear that the estimate (24) is monotonic in R'. Therefore, in equation (82) we may replace  $E_{\beta}(\xi)$  with any upper-bound of it. We will use the well-known upper-bound, which leads to binomial estimates of spectrum components [15, (46)]:

$$E_{\beta}(\xi_0) \le \frac{1}{2} (\log 2 + h(\beta) - h(\xi_0)).$$
 (86)

Furthermore, it can also be argued that maximum cannot be attained by  $\xi_0$  so small that

$$h(\beta) - \frac{1}{2}(\log 2 + h(\beta) - h(\xi_0)) < 0.$$

So from now on, we assume that

$$h^{-1}(\log 2 - h(\beta)) \le \xi_0 \le \delta$$

and that  $x = x(\xi_0) \leq 2\xi_0(1-\xi_0)$  is determined from the equation:

$$\log 2 - R = \xi_0 h\left(\frac{x}{2\xi_0}\right) + (1 - \xi_0) h\left(\frac{x}{2(1 - \xi_0)}\right)$$
(87)

(we remind  $R = h(\beta)$ ).

We proceed to demonstrating (83). For convenience, we introduce

$$a_1 \stackrel{\triangle}{=} 1 - \frac{x}{2\xi_0} \,, \tag{88}$$

$$a_2 \stackrel{\triangle}{=} \frac{x}{2 - 2\xi_0} \,. \tag{89}$$

By constraints on x it is easy to see that

$$0 \le a_2 \le \min(a_1, 1 - a_1)$$
.

Therefore, we have

$$\log 2 - R = \xi_0 h(a_1) + (1 - \xi_0) h(a_2) \ge h(a_2)$$

and thus  $a_2 \leq a_{max}$  defined in (84). Similarly, we have

$$\log 2 - R = \xi_0 h(a_1) + (1 - \xi_0) h(a_2) \le \xi_0 \log 2 + (1 - \xi_0) h(a_2)$$

and since  $\xi_0 \leq \delta$  we get that  $a_2 \geq a_{min}$  defined in (85). Next, notice that  $\frac{h(x)}{x(1-x)}$  is decreasing on (0, 1/2]. Thus, we have

$$h(a_1) \ge g_0(a_1) 4 \log 2$$
 (90)

$$h(a_{2}) \geq h(a_{max}) \frac{g_{0}(a_{2})}{g_{0}(a_{max})} = \frac{\log 2 - R}{a_{max}(1 - a_{max})} g_{0}(a_{2}) \stackrel{\triangle}{=} c \cdot g_{0}(a_{2}), \qquad (91)$$

where in the last step we introduced  $c > 4 \log 2$  for convenience. Consequently, we get

$$\log 2 - R$$

$$=\xi_0 h(a_1) + (1 - \xi_0) h(a_2)$$
(92)

$$\geq 4\log 2 \cdot \xi_0 g_0(a_1) + (1 - \xi_0)c \cdot g_0(a_2) \tag{93}$$

$$= 4\log 2 \cdot g_0(\xi_0, x) + (1 - \xi_0)(c - 4\log 2) \cdot g_0(a_2)$$
(94)

$$\geq 4\log 2 \cdot g_0(\xi_0, x) + (1 - \delta)(c - 4\log 2) \cdot g_0(a_{min}) \,. \tag{95}$$

Rearranging terms yield (83).

We proceed to proving monotonicity of (82). The technique we will use is general (can be applied to L > 3 and j > 1), so we will avoid particulars of L = 3, j = 1 case until the final step.

Notice that regardless of the function  $q(\nu)$  we have the equivalence:

$$\frac{d}{d\xi_0} \xi_0 g(a_1) + (1 - \xi_0) g(a_2) \ge 0 \quad \iff \\ \frac{1}{2} \frac{dx}{d\xi_0} (g'(a_2) - g'(a_1)) \ge \int_{a_2}^{a_1} (1 - x) (-g''(x)) dx - g'(a_2) ,$$
(96)

where we recall definition of  $a_1, a_2$  in (88)-(89). Differentiating (87) in  $\xi_0$  (and recalling that R is fixed, while  $x = x(\xi_0)$ is an implicit function of  $\xi_0$ ) we find

$$\frac{dx}{d\xi_0} = -2\frac{\log\frac{1-a_2}{a_1}}{\log\frac{1-a_2}{a_2}\frac{a_1}{1-a_1}} < 0$$

Next, one can notice that the map  $(\xi_0, x, R) \mapsto (a_1, a_2)$  is a bijection onto the region

$$\{(a_1, a_2) : 0 \le a_1 \le 1, 0 \le a_2 \le a_1(1 - a_1)\}.$$
 (97)

With the inverse map given by

$$\xi_0 = \frac{a_2}{1 - a_1 + a_2}, x = \frac{2a_2^2}{1 - a_1 + a_2},$$
  
$$R = \log 2 - \xi_0 h(a_1) - (1 - \xi_0) h(a_2).$$

Thus, verifying (96) can as well be done for all  $a_1, a_2$  inside the region (97). Substituting  $g = g_1$  into (96) we get that monotonicity in (82) is equivalent to a two-dimensional inequality:

$$-2\log\frac{1-a_2}{a_1} \cdot (a_1^2 - a_2^2)$$
  

$$\geq (2a_1^2 - \frac{4}{3}(a_1^3 - a_2^3) - 1)\log\frac{1-a_2}{a-2}\frac{a_1}{1-a_1}.$$
 (98)

It is possible to verify numerically that indeed (98) holds on the set (97). For example, one may first demonstrate that it is sufficient to restrict to  $a_2 = 0$  and then verify a corresponding inequality in  $a_1$  only. We omit mechanical details.

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