

# Graph entropy and the zero-error capacity of channels

Jaikumar Radhakrishnan



Tata Institute of Fundamental Research, Mumbai

Based on joint work with Siddharth Bhandari (TIFR),  
Venkat Guruswami (CMU) and Marco Dalai (Brescia)

# The binary difference problem

Fix  $N$ . What is the minimum  $n$  such that we can find  $N$  **different** sequences in  $\{0, 1\}^n$ ?

**Answer:**  $\lceil \log_2 N \rceil$

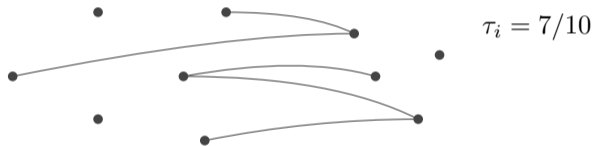
What is the maximum number of sequences in  $\{0, 1\}^n$  such that every **two** of them are **different**?

**Answer:**  $2^n$

# Hansel's lemma about graph covering

Suppose  $[M] = \{1, 2, \dots, M\}$ ,  $K_M$  = the complete graph on  $[M]$ , and

- $G_i$  ( $i \in I$ ) a finite sequence of bipartite graphs on  $[M]$
- $\tau_i$  fraction of non-isolated vertices in  $G_i$



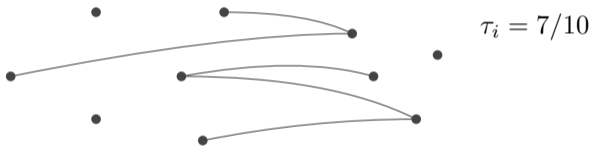
Lemma (Hansel '64)

$$\bigcup_{i \in I} G_i = K_M \implies \sum_{i \in I} \tau_i \geq \log_2(M).$$

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# Entropy of a random variable

## Definition (Shannon entropy)



Claude Shannon (1916-2001)

Let  $\mathbf{X}$  be a random variable taking values in the set  $[n] = \{1, 2, \dots, n\}$ . Then, its entropy is given by

$$H[\mathbf{X}] = - \sum_{i=1}^n p_i \log_2 p_i.$$

- $H[\mathbf{X}]$  measures the uncertainty in  $\mathbf{X}$ .
- $H[\mathbf{X}]$  is a function of the distribution of  $\mathbf{X}$ , not the actual values it takes.
- $H[\mathbf{X}] \leq \log_2 n$ .

# Pragmatics

Alice

Observes  $\mathbf{X}$

Sends Bob a message  $\mathbf{M}$

$\Leftarrow$

Bob

$\Rightarrow$

Recovers  $\mathbf{X}$  from  $\mathbf{M}$

Goal

- Alice and Bob exchange bits.
- Bob must recover  $\mathbf{X}$  exactly.
- Goal: minimize the (expected) total number of bits transmitted.

Transmission cost

Let  $T[\mathbf{X}]$  denote the minimum (expected) cost of transmitting  $\mathbf{X}$ .

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# Entropy and transmission

## Theorem

$$H[\mathbf{X}] \leq T[\mathbf{X}] \leq H[\mathbf{X}] + 1.$$



# Long years ago ... 1948

## Shannon's source coding theorem

Let  $p$  be a probability distribution on  $[n]$ . For  $\epsilon > 0$  and positive integer  $k$ , let

$$N(k, \epsilon) = \min_{A \subset [n]^k: p^k(A) \geq 1 - \epsilon} |A|.$$

## Theorem (Shannon)

For all  $\epsilon \in (0, 1)$ ,  $\lim_{k \rightarrow \infty} \frac{1}{k} \log_2 |N(k, \epsilon)| = H(p)$ .

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... not wholly or in full measure, but very substantially!

# Conditional entropy

## Definition

$(\mathbf{X}, \mathbf{Y})$ : a pair of random variables with some joint distribution.

$$H[\mathbf{Y} | \mathbf{X}] = \sum_i p_X(i) H[\mathbf{Y}_i]$$

## Fact

- *Conditioning reduces uncertainty:  $H[\mathbf{Y} | \mathbf{X}] \leq H[\mathbf{Y}]$ .*
- $H[\mathbf{XY}] = H[\mathbf{X}] + H[\mathbf{Y} | \mathbf{X}]$ .

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# The noisy channel

## Specification

Input alphabet:  $[m]$

Output alphabet:  $[n]$

Characteristics:  $\Pr[\text{output} = j \mid \text{input} = i] = p_{j|i}$ .

## Code of conduct

Encoding:  $\{0, 1\}^k \rightarrow [m]^t$

Decoding:  $[n]^t \rightarrow \{0, 1\}^k$

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Error:  $\Pr[\text{input} \neq \text{output}] \leq \epsilon$ .

Rate:  $\frac{k}{t}$  should be as large as possible.

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# Capacity

Input to the channel:  $\mathbf{X} \in [m]$

Output of the channel:  $\mathbf{Y} \in [n]$ .

Definition (Capacity of a channel  $E$ )

$$C(E) = \max_{\mathbf{X}} H[\mathbf{X}] + H[\mathbf{Y}] - H[\mathbf{XY}].$$

# Jaane kya toone kahi . . . jaane kya meine suni

## Theorem (Shannon)

*Let  $C$  be the capacity of the channel. Then, for all  $\epsilon > 0$  and all  $k$ , there exist encoders and decoders such that*

**Encoding rate:**  $\frac{k}{t} \geq C - \epsilon$ .

**Error:**  $\Pr[\text{error}] \rightarrow 0$  as  $k \rightarrow \infty$ .

**Optimality:** *Can't replace  $C - \epsilon$  by  $C + \delta$  for any  $\delta > 0$ .*

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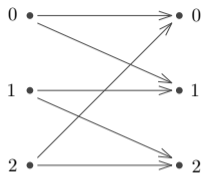
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# The 3/2 channel: trifference



- Any two input symbols are confusable (share a common output).
  - Any two codewords are confusable (in any position).
  - Zero-error capacity 0.

- However, the three inputs have no common output.

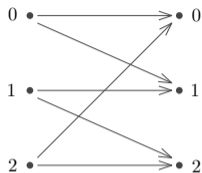
■ We can build a series of channels which cannot be identified (they can be distinguished) but we can narrow down the intended codeword to a list of size  $L=2$ .

$$Y = 0 \rightarrow \{0, 2\} \quad 1 \rightarrow 2$$

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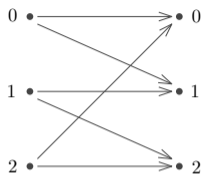
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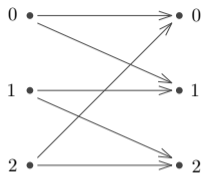
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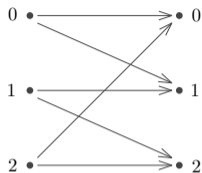
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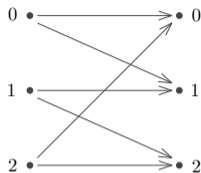


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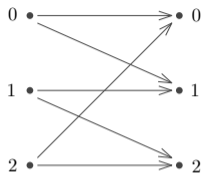
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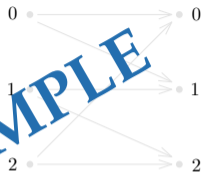
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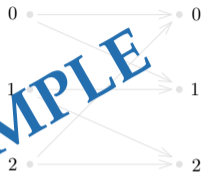


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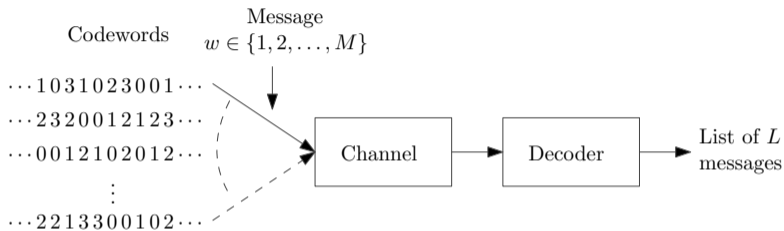
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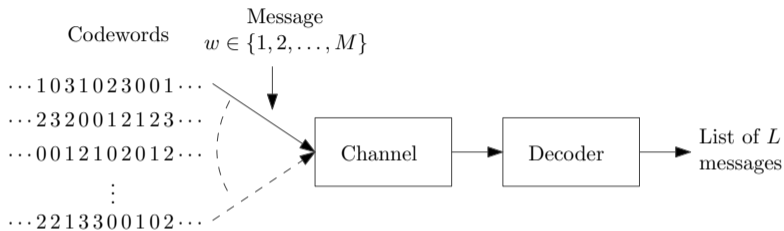
$x$	=	0	0	0	2	1	2	...
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# Zero-error capacity with list decoding



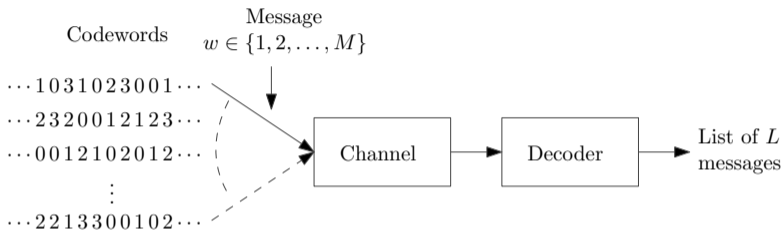
- The decoder outputs a *list* of  $L$  messages
- Zero-error code: the correct message is always in the list
  - ⇔ No  $L + 1$  codewords are compatible with any output sequence
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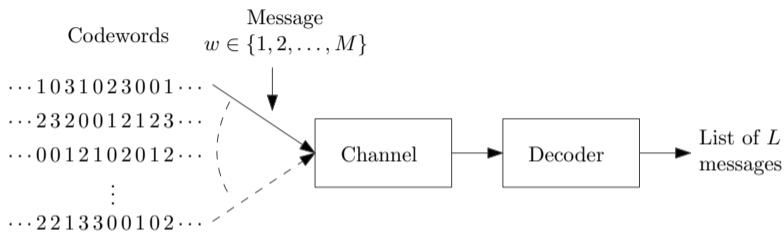
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# Ternary encoding

Let  $R_3(n)$  be the size of the largest subset  $C \subseteq \{0, 1, 2\}^n$  such that every **three** of them are **trifferent**?

**trifferent**  $\equiv \forall$  (distinct)  $x, y, z \in C \quad \exists i : \{x_i, y_i, z_i\} = \{0, 1, 2\}$

$$\begin{array}{rcccccccc} x & = & 0 & 0 & 0 & 2 & 1 & 2 & \dots \\ y & = & 0 & 1 & 1 & 1 & 2 & 1 & \dots \\ z & = & 1 & 1 & 2 & 2 & 1 & 2 & \dots \end{array}$$

**Answer:**  $2^{0.212n} \leq R_3(n) \leq 2 \left(\frac{3}{2}\right)^n \approx 2^{0.545n}$  (Körner & Marton '88)

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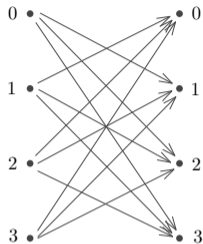
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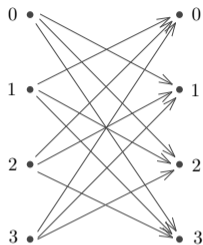
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• The four inputs have no common output

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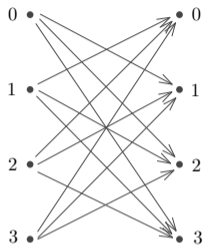
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⇒ We can build 4-tuples of codewords which cannot be confused

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... Upper bounds for zero-error capacity with  $L = 3$ ?

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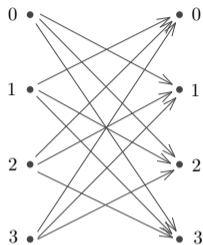
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## Problem

Let  $R_4(n)$  be the size of the largest code  $C \subseteq \{0, 1, 2, 3\}^n$  that satisfies

$$\forall (\text{distinct}) w, x, y, z \in C \quad \exists i : \{w_i, x_i, y_i, z_i\} = \{0, 1, 2, 3\}$$

- Standard random coding:  $R_4(n) \geq 2^{0.0473n}$
- Simple upper bound  $R_4(n) \leq 3 \cdot \left(\frac{4}{3}\right)^n \approx 2^{0.415n}$
- Fredman-Komlós '84:  $R_4(n) \leq 2^{\frac{3}{8}(1+o(1))n} \approx 2^{0.375n}$
- Arikan '94:  $R_4(n) \leq 2^{0.3512(1+o(1))n}$ , using Plotkin the bound can be improved to about  $2^{0.35n}$
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$$\forall (\text{distinct}) w, x, y, z \in C \quad \exists i : \{w_i, x_i, y_i, z_i\} = \{0, 1, 2, 3\}$$

- Standard random coding:  $R_4(n) \geq 2^{0.0473n}$
- Simple upper bound  $R_4(n) \leq 3 \cdot \left(\frac{4}{3}\right)^n \approx 2^{0.415n}$
- Fredman-Komlós '84:  $R_4(n) \leq 2^{\frac{3}{8}(1+o(1))n} \approx 2^{0.375n}$
- Arikan '94:  $R_4(n) \leq 2^{0.3512(1+o(1))n}$ , using Plotkin the bound can be improved to about  $2^{0.3276n}$ , using the linear programming bound

# Alphabet size 4

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$$R_4(n) \leq 2^{\frac{6}{19}(1+o(1))n} \approx 2^{0.3158n},$$

that is,

$$\log \frac{R_4(n)}{n} \leq \frac{6}{19}(1 + o(1)) \approx 0.3158.$$

(Joint work with Venkat Guruswami and Marco Dalai '17)

# The framework (due to Fredman and Komlós)

Let  $C \subseteq \{0, 1, 2, 3\}^n$  be a 4-hash code.

**Step 1:** Fix distinct  $x, y \in C$ .

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$\implies$  if  $x_i = y_i$ , set  $z_i$  to  $*$ ;

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# The picture

$$\begin{bmatrix} 3 & 1 & 3 & 0 & 0 & 1 & \dots \\ 0 & 1 & 1 & 2 & 3 & 0 & \dots \\ 3 & 2 & 2 & 2 & 0 & 3 & \dots \\ 2 & 3 & 1 & 3 & 2 & 2 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \dots \\ 1 & 3 & 1 & 1 & 0 & 1 & \dots \end{bmatrix}$$

$\Rightarrow$

$$\begin{bmatrix} 3 & 1 & 3 & 0 & 0 & 1 & \dots \\ 0 & 1 & 1 & 2 & 3 & 0 & \dots \\ * & * & 1 & * & * & 1 & \dots \\ 1 & * & * & 1 & 1 & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \dots \\ 0 & * & * & 0 & * & * & \dots \end{bmatrix}$$

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# The Fredman-Komlós bound

Recall that  $C \subseteq \{0, 1, 2, 3\}^n$  be a 4-hash code.

- Pick distinct  $x, y \in C$  uniformly at random.
- Let  $f[0], f[1], f[2], f[3]$  be the frequencies of the symbols in  $i$ -th coordinate. Then,

$$\mathbb{E}[\tau(i)] \approx \sum_{a \neq b} f[a] f[b] (1 - f[a] - f[b]) \leq \frac{3}{8}.$$

- Use Hansel's lemma to obtain

$$\log_2(|C| - 2) \leq \sum_i \tau(i) \leq \left(\frac{3}{8}\right) n.$$

- Conclusion:  $|C| \leq 2^{((3/8)+o(1))n}$ .

# The Arikan bound

- Pick two distinct  $x, y \in \mathcal{C}$  that agree on many coordinates.  
(Coding theory says: high rate  $\implies$  small distance, say  $\Delta n$ .)
- Use an *ad hoc* argument to argue that all four symbols appear in each coordinate with frequency at least  $f_0$  (otherwise, the code must be small anyway).
- Use Hansel's lemma to conclude that

$$\log_2(|\mathcal{C}| - 2) \leq \sum_i \tau(i) \leq \Delta(1 - 2f_0)n.$$

- Using bounds on distance in terms of rate gives  $|\mathcal{C}| \leq 2^{0.3512n}$  (using the Plotkin bound) and  $|\mathcal{C}| \leq 2^{0.3276n}$  (using the linear programming bound).



# Looking deeper into the Plotkin bound

Recall that  $C \subseteq \{0, 1, 2, 3\}^n$  is a 4-hash code.

## Main idea

Generate  $x, y$  more carefully, in a way reminiscent of the proof of the Plotkin bound.

- Fix  $s \leq n$ . Later we will set  $s \approx \left(\frac{1}{2}\right) \log_2 |C|$ .
- Pick  $x \in C$  uniformly at random; let  $w = x[1, \dots, s]$  be the prefix of  $x$  of length  $s$ .
- Pick  $y \in C$  distinct from  $x$  uniformly at random conditioned on  $y[1, \dots, s] = w$ .
- This helps us derive a stronger conclusion from Hansel's lemma.

# Analysis

- $\tau(i) = 0$  if  $i = 1, 2, \dots, s$ .
- For  $i = s + 1, s + 2, \dots, n$ , the quantity  $\tau(i)$  is a random variable depending on our choice of  $x$  and  $y$ . We will show that  $\mathbb{E}[\tau(i)] \leq \frac{3}{8}$ .
- Use Hansel's lemma to conclude that

$$\log_2(|C| - 2) \leq \left(\frac{3}{8}\right) (n - s).$$

- Set  $s \approx \left(\frac{1}{2}\right) \log_2 |C|$  and obtain

$$|C| \leq 2^{(6/19)n}.$$

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# In expectation

For each prefix  $w \in \{0, 1, 2, 3\}^s$ , let  $C_w$  be the subcode

$$C_w = \{x \in C : x[1, \dots, s] = w\}.$$

For  $i > s$ , let  $f_{i,w}[a]$  be the frequency of the symbol  $a$  in the  $i$ -th coordinate in the subcode  $C_w$ ; let  $\tau(i|w) = \sum_{a \neq b} f_{i,w}[a]f_{i,w}[b](1 - f_i[a] - f_i[b])$ . Then,

$$\mathbb{E}[f_{i,w}] = f_i \quad \text{and} \quad E[\tau(i)] = \mathbb{E}_W[\phi(f_{i,w}, f_i)],$$

where  $\phi(f', f) = \sum_{a \neq b} f'[a]f'[b](1 - f[a] - f[b])$ .

## Claim

*Suppose  $\mathbb{E}[f_w] = f$  and each component of  $f$  is at most  $\frac{1}{2}$ . Then,  $\mathbb{E}_W[\phi(f_w, f)] \leq \phi(f, f) \leq \frac{3}{8}$ .*

# Must $\phi$ meet our expectation?

## Claim

Suppose  $\mathbb{E}[f_W] = f$  and each component of  $f$  is at most  $\frac{1}{2}$ . Then,

$$\mathbb{E}_W[\phi(f_W, f)] \leq \phi(f, f) \leq \frac{3}{8}.$$

- Let  $\Delta_w = f_w - f$ ; then,  $\mathbb{E}_W[\phi(f_W, f)] = \phi(f, f) - \mathbb{E}_W[\Delta_w^t N \Delta_w]$ , where

$$N = - \begin{pmatrix} 0 & f[2] + f[3] & f[1] + f[3] & f[1] + f[2] \\ f[2] + f[3] & 0 & f[0] + f[3] & f[0] + f[2] \\ f[1] + f[3] & f[0] + f[3] & 0 & f[0] + f[1] \\ f[1] + f[2] & f[0] + f[2] & f[0] + f[1] & 0 \end{pmatrix}.$$

- Now,  $\mathbf{1} \cdot \Delta_w = 0$ . In the subspace orthogonal to  $\mathbf{1}$ , the matrix  $N$  is positive semidefinite (because of our assumption on  $f$ ).

# What about the $q/(q - 1)$ channel?

Let  $R_{q,L}(n)$  be the size of the largest subset  $C \subseteq \{0, 1, 2, \dots, q - 1\}^n$  such that every  $L + 1$  of them are **qifferent**?

**qifferent**  $\equiv \forall$  (distinct)  $x_1, x_2, \dots, x_{L+1} \in C \quad \exists i \in [L + 1] :$   
 $\{x_1[i], x_2[i], \dots, x_{L+1}[i]\} = \{0, 1, 2, \dots, q - 1\}.$

Theorem (with Siddharth Bhandari '18)

For every  $\epsilon < 1/6$ , for all large  $q$ , if  $L \leq \epsilon q \ln q$ , then

$$\frac{\log R_{q,L}(n)}{n} = O(\exp(-q^{1-6\epsilon}/8)).$$

Earlier, it was known that the rate of the code is exponentially small if the list size is bounded by  $1.58q$  (Chakraborty, Radhakrishnan, Raghunathan and Sasatte '06).



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