

# The Unreasonable Effectiveness

## of Expanders in Coding Theory

Fernando Granha Jeronimo  
(UIUC)

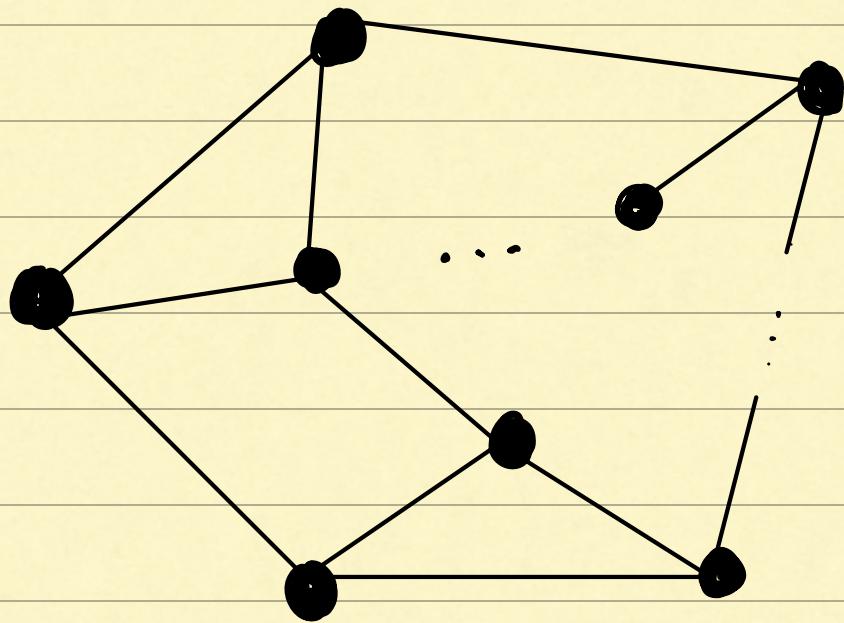
# A Glimpse of The Unreasonable Effectiveness of Expanders in Coding Theory

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(UIUC)

What is one widely studied object  
in CS/Combinatorics?

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in CS/Combinatorics?

Graphs  $G = (V, E)$



# Expander Graphs

Combine two appealing properties:

Well-connected yet sparse

- Mimic the complete graph ✓
- Sparse: can have  $O(1)$  degree ✓
- Have (pseudo) random properties ✓
- Can be constructed explicitly ✓

# Expander Graphs

Combine two appealing properties:

Well-connected yet sparse

I work with algorithm design.

Why should I care?

# Expander Graphs

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Well-connected yet sparse

I work with algorithm design.

Why should I care?

There is an ongoing revolution  
in the design of fast algorithms

Structure - VS - Randomness:

Take an arbitrary  $G = (V, E)$  and

decompose it into expanding pieces



# Expander Graphs

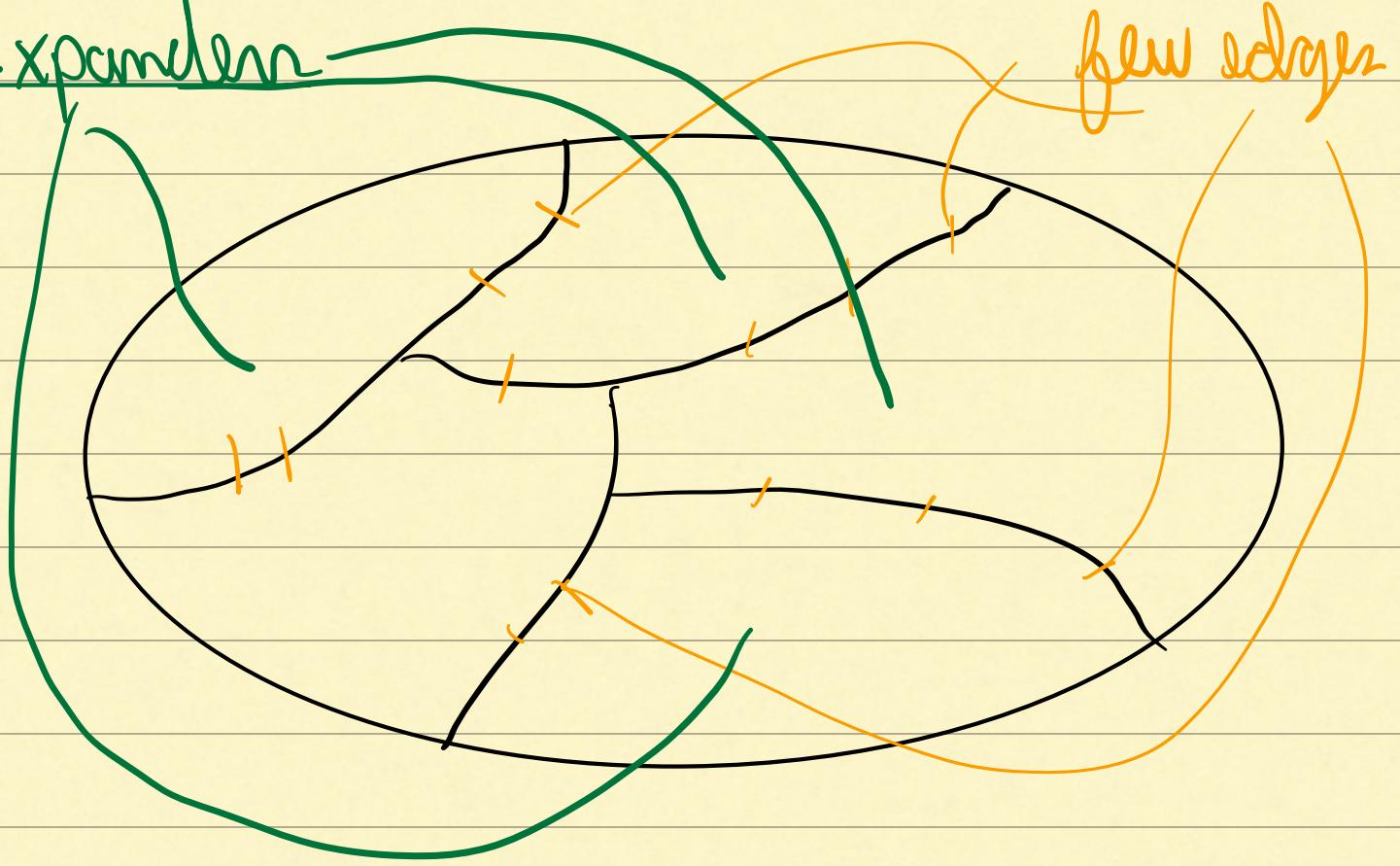
There is an ongoing revolution  
in the design of fast algorithms

Structure -vs- Randomness:

Take an arbitrary  $G = (V, E)$  and

decompose it into expanding pieces

Expanders



# Expander Graphs

Applications to:

- Algorithm Design
  - Complexity Theory (PCPs / Hardness)
  - Derandomization
  - Coding Theory
  - MCMC (Mixing / Sampling / Counting)
  - Group and Number Theory
  - Combinatorics
- ⋮

... graphs do seem fundamental.

What about the study of strings?

$\Sigma$  finite alphabet

$$C \subseteq \Sigma^n$$

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Code

Coding Theory: The study of strings

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↗  
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Coding Theory: The study of strings

Applications to:

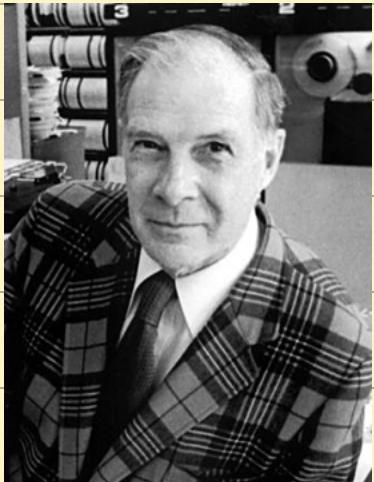
- Communication / Storage
  - Complexity Theory / Pseudorandomness
  - Quantum Computing / Advantage
- ⋮

# Coding Theory

Two Seminal Models ~1940-1950

Hamming Model

Errors are adversarial



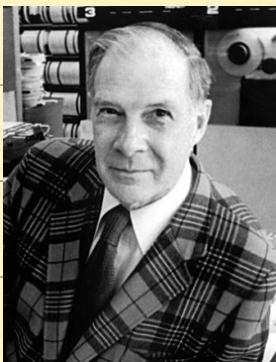
Shannon Model

Errors are Probabilistic



# Coding Theory

## Hamming Model



Errors are adversarial

$$\mathcal{C} \subseteq \Sigma^n$$

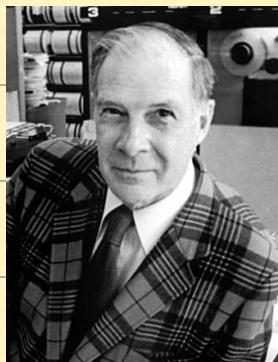
$$\Delta(\mathcal{C}) = \min_{x, y \in \mathcal{C}: x \neq y} \Delta(x, y) \in [0, 1] \text{ (distance)}$$

$$r(\mathcal{C}) = \frac{\log_{|\Sigma|}(|\mathcal{C}|)}{n} \in [0, 1] \text{ (rate)}$$

# Coding Theory

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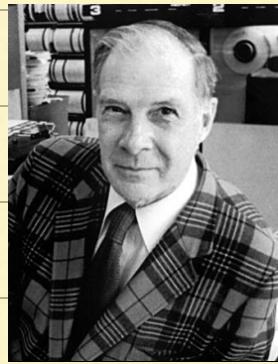
$$r(\mathcal{C}) = \frac{\log_{|\Sigma|}(|\mathcal{C}|)}{n} \in [0, 1] \text{ (rate)}$$

We want both  $\Delta(\mathcal{C})$  and  $r(\mathcal{C})$  large, but there is tension

# Coding Theory

## Hamming Model

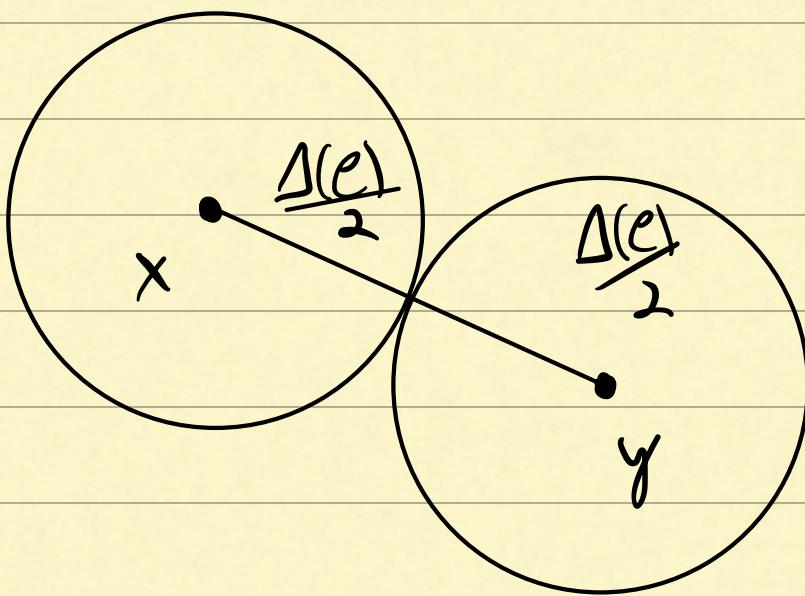
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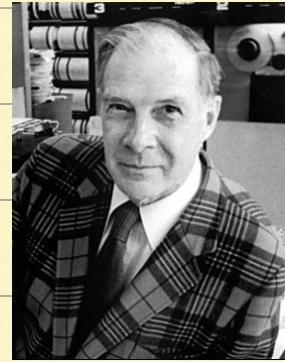
$$e \subseteq \Sigma^n$$

$$\Delta(e) = \min_{x, y \in \mathcal{C}, x \neq y} \Delta(x, y) \in [0, 1] \text{ (distance)}$$

$$r(e) = \frac{\log_{121}(|\mathcal{C}|)}{n} \in [0, 1] \text{ (rate)}$$



# Coding Theory



## Hamming Model

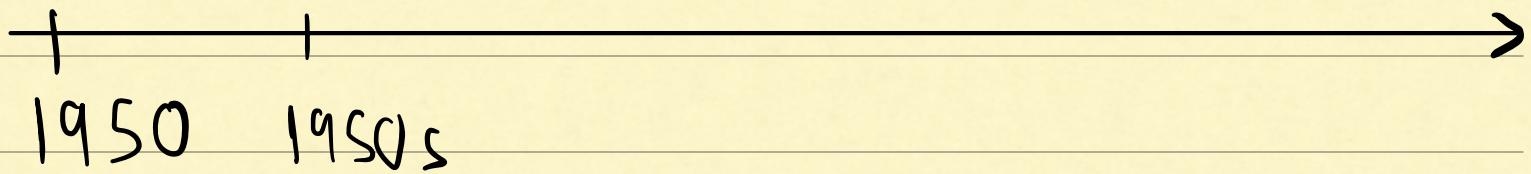
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$$e \subseteq \Sigma^n$$

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## Hamming model

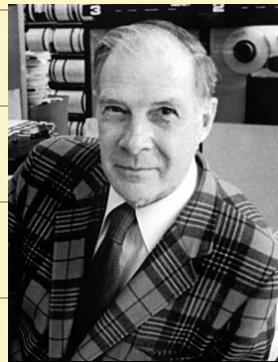


Random codes  
Gilbert - Varshamov

# Coding Theory

## Hamming Model

Errors are adversarial



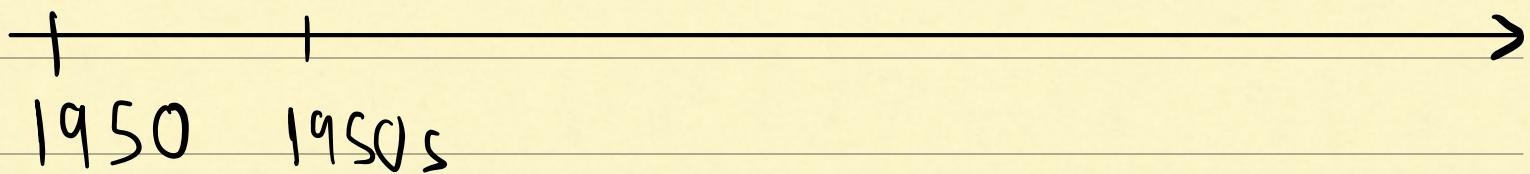
$$C \subseteq \Sigma^n$$

$$\Delta(C) = \min_{x, y \in C: x \neq y} \Delta(x, y) \in [0, 1] \text{ (distance)}$$

$$r(C) = \frac{\log_2(|C|)}{n} \in [0, 1] \text{ (rate)}$$

How to construct explicit codes?

Hamming  
model

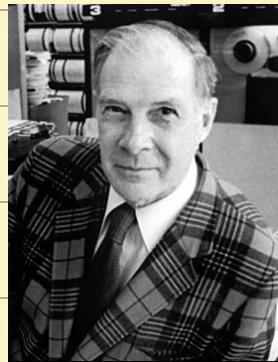


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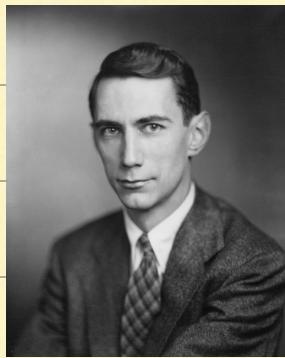
Ta-Shma's  
codes



Random codes  
Gilbert - Varshamov

# Coding Theory

## Shannon Model



Errors are Probabilistic

Ex binary symmetric channel  $BSC_p$

Each bit is independently flipped w.p.  $p \in [0, 1]$

Using Probabilistic Method Shannon proved

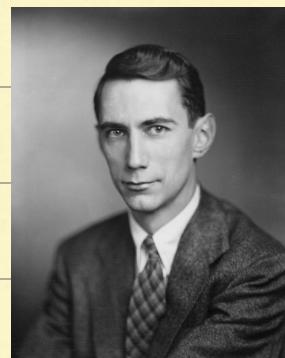
Theorem [Capacity]  $\forall \epsilon > 0, p < 1 - \frac{1}{q}$

Rate  $\geq 1 - H_q(p) - \epsilon$  (but no more)

# Coding Theory

## Shannon Model

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How to construct explicit capacity achieving code?

Shannon's

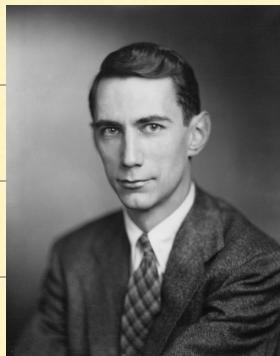
random codes

| →

1948

# Coding Theory

## Shannon Model



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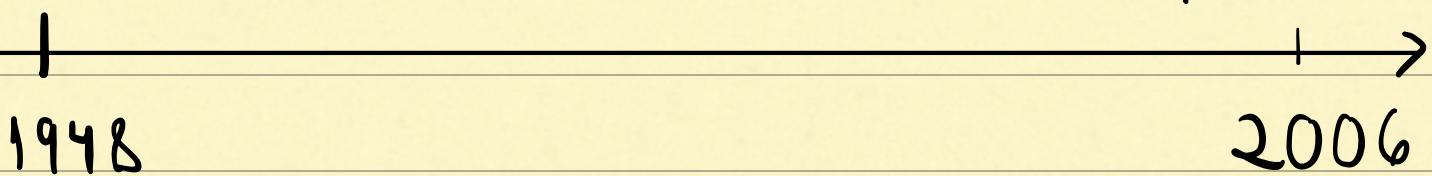
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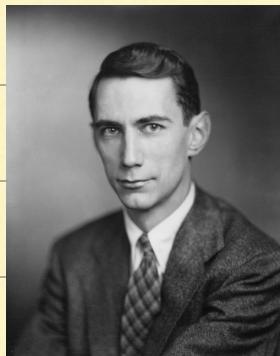
Shannon's  
random codes

Arikan's  
polar codes



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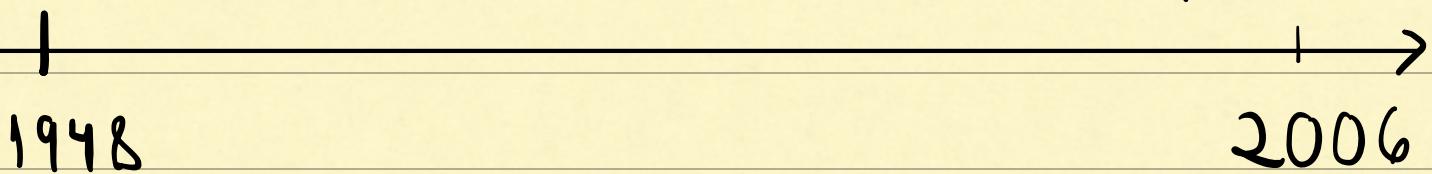
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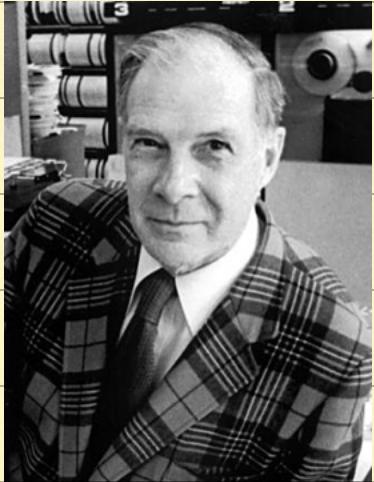


# Coding Theory

Two Seminal Models ~1940-1950

Hamming Model

Errors are adversarial



Shannon Model

Errors are Probabilistic



More general errors ✓

Simpler errors ↗

Worse Parameters ↗

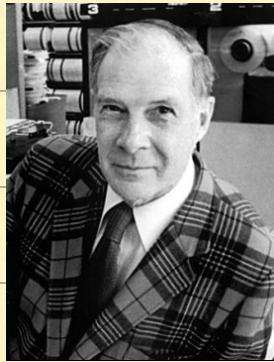
Better Parameters ✓

# Coding Theory

## Hamming Model

Errors are adversarial

$$C \subseteq \sum^n \quad p = \Delta(e)$$



How can we decode from  $\geq \frac{1}{2}$

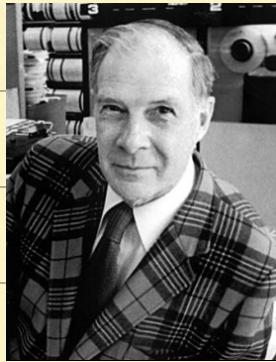
adversarial errors?

# Coding Theory

## Hamming Model

Errors are adversarial

$$C \subseteq \sum^n \quad p = \Delta(e)$$



How can we decode from  $\geq \frac{r}{2}$  adversarial errors?

Impossible to unique decode!

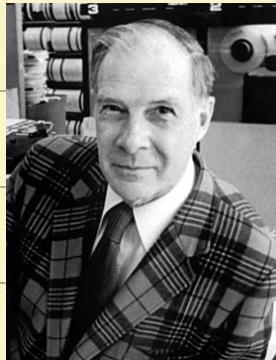
Possible if we allow a list of codewords

[Elias '57, Wagstaff '58]

# Coding Theory

## Hamming Model

Errors are adversarial

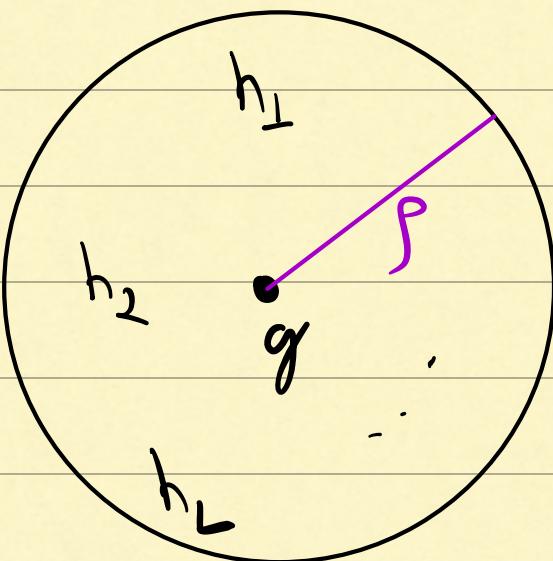


$$e \subseteq \Sigma^n$$

## List Decoding [Elias '59, Wozencraft '58]

$e$  is  $(p, L)$ -list decodable if

$$\forall g \in \Sigma^n, |\text{Ball}(g, p) \cap e| \leq L.$$

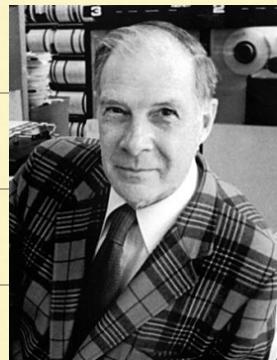


# Coding Theory

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Zyablov-Pinsker '81 via Probabilistic Method

Theorem [List Decoding Capacity]

$$\forall \epsilon > 0, p \in (0, 1 - \frac{1}{q})$$

(1)  $\exists C \subseteq F_q^n, (p, L)$ -list decodable

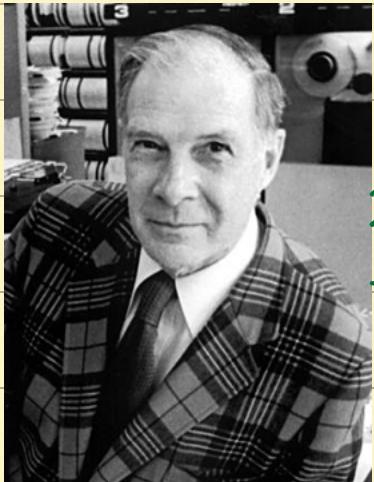
(with  $r(C) \geq 1 - H_q(p) - \epsilon$  and  $L = O\left(\frac{1}{\epsilon}\right)$ )  
(best possible)

# Coding Theory

Two Seminal Models ~1940-1950

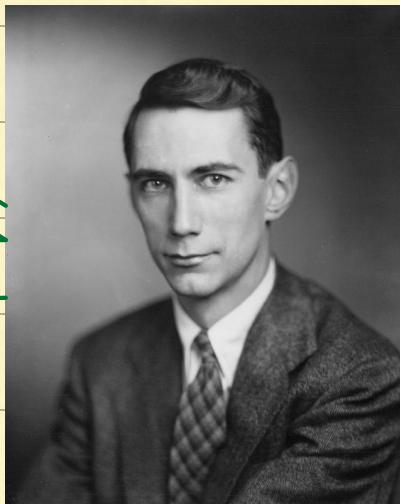
Hamming Model

Errors are adversarial



Shannon Model

Errors are Probabilistic



list decoding

More general errors ✓

Simpler errors ↗

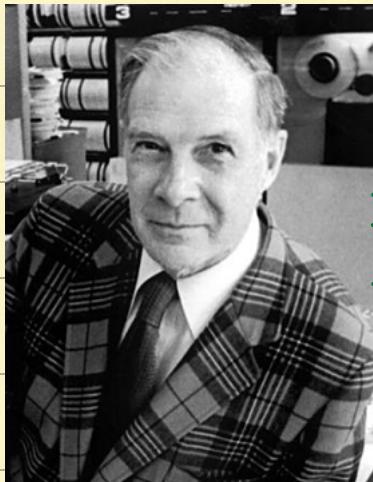
Worse Parameters ↗

Better Parameters ✓

# Chart to Construct Explicit Codes achieving List Decoding Capacity

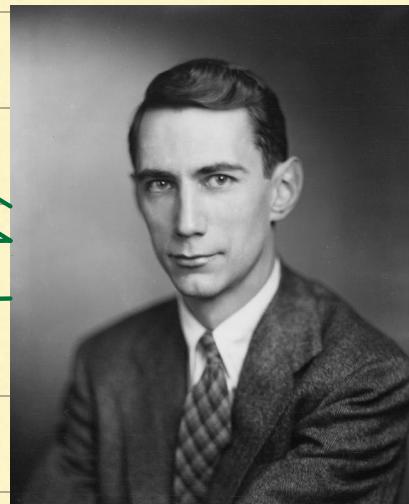
Hamming Model

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Shannon Model

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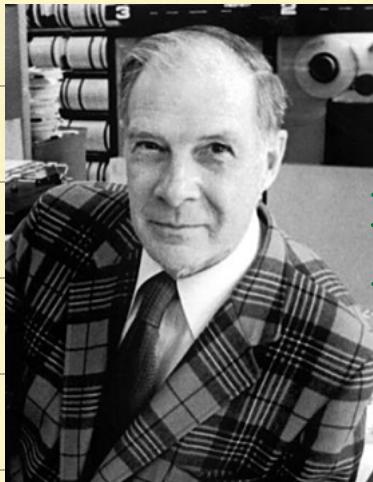
[Pernice - Sprumont - Wootters '25]

list decoding  
capacity  $\xrightarrow{*}$  capacity

# Chart to Construct Explicit Codes achieving List Decoding Capacity

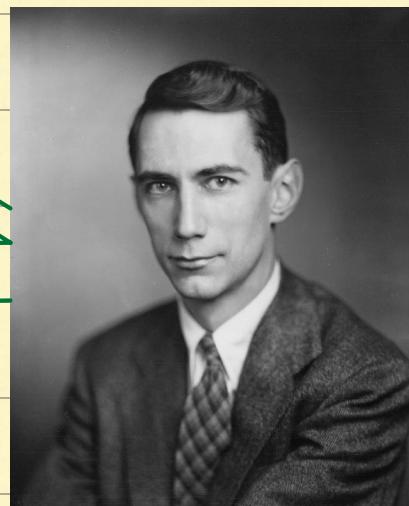
Hamming Model

Errors are adversarial



Shannon Model

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[Pernice - Sprumont - Wootters '25]

list decoding  
capacity  $\Rightarrow^*$  capacity



not always true

## Singleton Bound

Theorem [1950's]

If code  $C$  with  $r(C) = R$ , we have

$$\Delta(C) \leq 1 - R$$

Reed-Solomon codes achieve this bound  
(RS)

$\alpha_1, \alpha_2, \dots, \alpha_n \in F$  distinct

$$p(x) = c_0 + c_1 x + \dots + c_{K-1} x^{K-1}$$

$\downarrow$  Encoding map

$$(p(\alpha_1), p(\alpha_2), \dots, p(\alpha_n))$$

codeword

## Singleton Bound

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codeword

Drawback:  $|\mathbb{F}| \geq n$   
large alphabet

# List Decoding with Explicit Codes

Efficient list decoding for RS

$$1 - \sqrt{2R}$$

[Sudan '95]

$$1 - \sqrt{R}$$

[Guruswami-Sudan '97]

(Johnson bound)

Passarelli-Vardy '05

$$1 - O(R \log \frac{1}{R})$$

Folded Reed-Solomon (FRS)

$$1 - R - \epsilon$$

[Guruswami-Rudra '06]

(List decoding Capacity)

# Amazing Synergy

Expander + Codes

Recent breakthroughs:

Ta-Shma's Codes '17

Explicit Binary Codes  
near GV

[+60 years after  
random construction]

$C^3$ -LTC

[Dimov et al. '22)

Good Quantum LDPC Codes

[Panteleev - Kalachov '22]

## Warm-up: Code Concatenation

$$C_{\text{out}} \subseteq \Sigma_{\text{out}}^n$$

$$C_{\text{in}} \subseteq \Sigma_{\text{in}}^d \quad \text{with} \quad |C_{\text{in}}| = |\Sigma_{\text{out}}|$$

$$C_{\text{in}} \simeq \Sigma_{\text{out}}$$

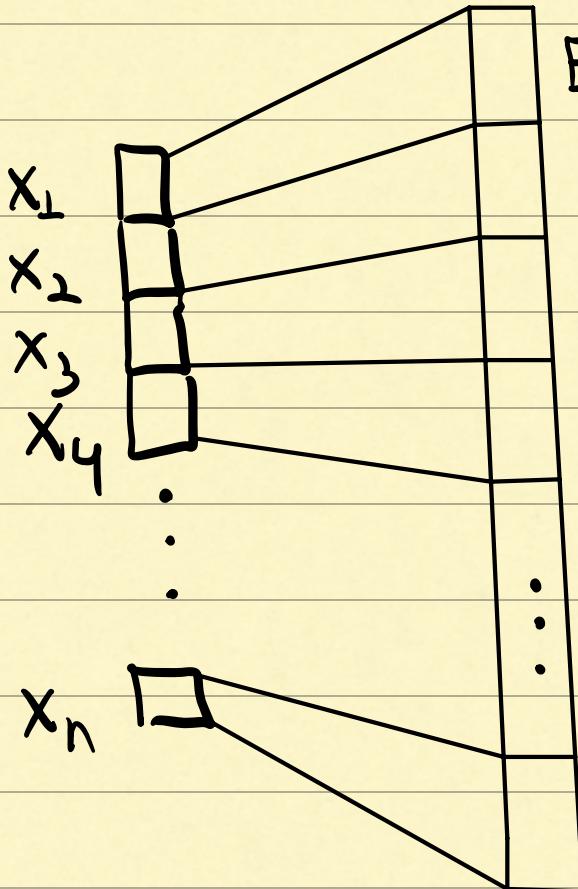
# Code Concatenation

$$x \in \mathcal{C}_{\text{out}} \subseteq \Sigma_{\text{out}}^n$$

$$\mathcal{C}_{\text{in}} \subseteq \Sigma_{\text{in}}^d \quad \text{with} \quad |\mathcal{C}_{\text{in}}| = |\Sigma_{\text{out}}|$$

$$\mathcal{C}_{\text{in}} \cong \Sigma_{\text{out}}$$

$$x = (x_1, x_2, \dots, x_n)$$



$$\text{Enc}_{\mathcal{C}_{\text{in}}}(x_1) \in \Sigma_{\text{in}}^d$$

$$\text{Enc}_{\mathcal{C}_{\text{in}}}(x_2) \in \Sigma_{\text{in}}^d$$

$$\text{Enc}_{\mathcal{C}_{\text{in}}}(x_3) \in \Sigma_{\text{in}}^d$$

$$\text{Enc}_{\mathcal{C}_{\text{in}}}(x_4) \in \Sigma_{\text{in}}^d$$

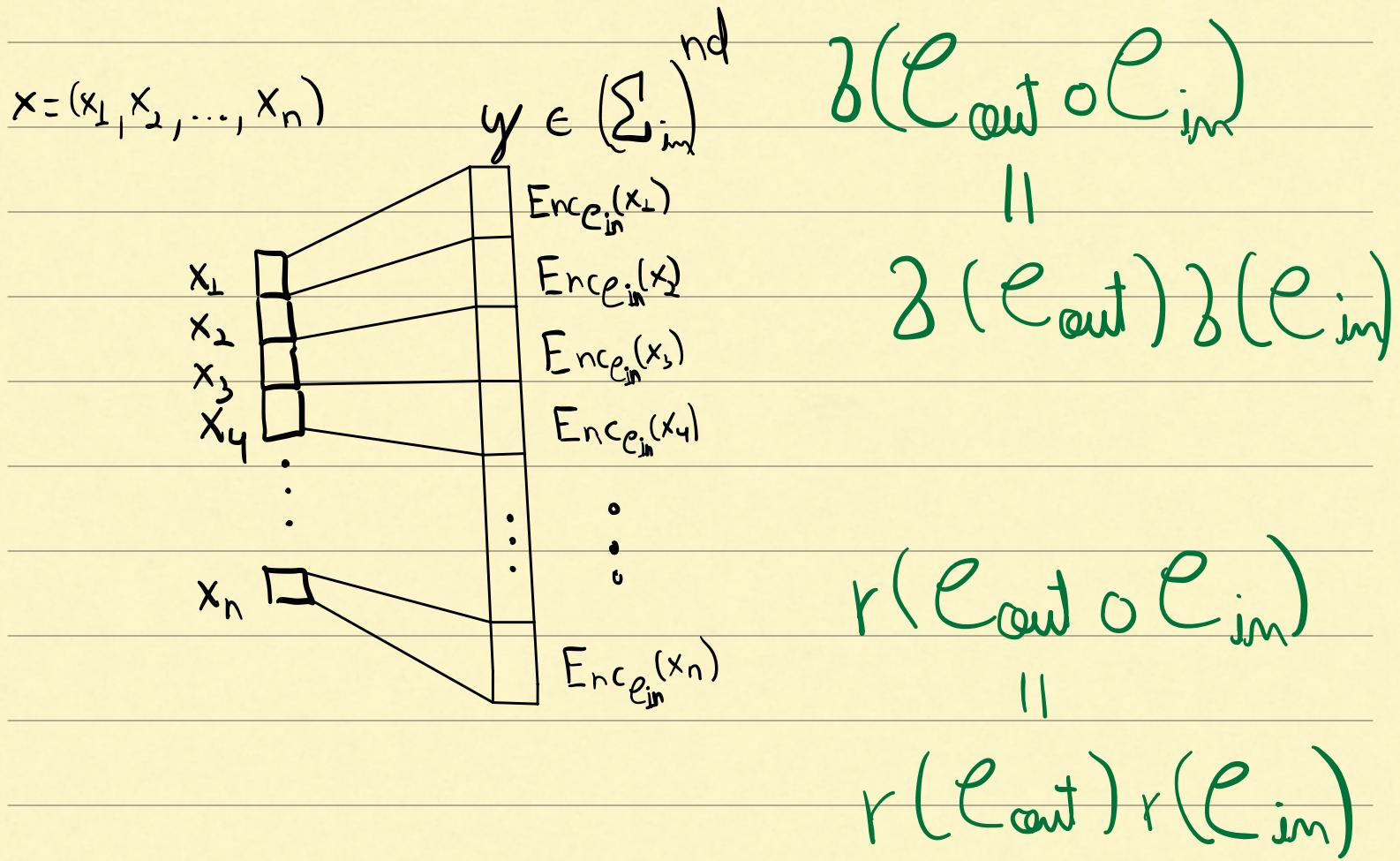
⋮

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## Code Concatenation

$$\delta(e_{\text{out}} \circ e_{\text{in}}) = \delta(e_{\text{out}}) \delta(e_{\text{in}})$$

$$r(e_{\text{out}} \circ e_{\text{in}}) = r(e_{\text{out}}) r(e_{\text{in}})$$

Dream Goal: Another code operation  $\square$

$$\delta(e_{\text{out}} \square e_{\text{in}}) \approx \delta(e_{\text{in}})$$

$$r(e_{\text{out}} \square e_{\text{in}}) \approx r(e_{\text{in}})$$

The "good" properties of a small local code  $e_{\text{in}}$  become the properties of the global code  $(e_{\text{out}} \square e_{\text{in}})$

## Local-to-Global

Dream Goal: Another code operation  $\square$

$$\delta(e_{\text{out}} \square e_{\text{in}}) \approx \delta(e_{\text{in}})$$

$$r(e_{\text{out}} \square e_{\text{in}}) \approx r(e_{\text{in}})$$

The "good" properties of a small local code  $e_{\text{in}}$   
become the properties of the global code  $(e_{\text{out}} \square e_{\text{in}})$

[Ahn - Edmonds - Luby '94]

This is possible using expander graphs!

AEL Construction

## AEL Construction

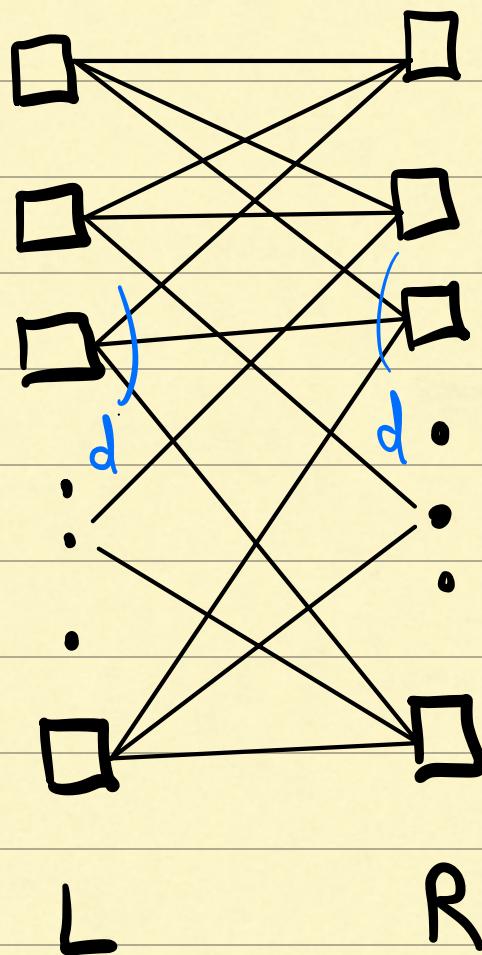
$AEL(G, \mathcal{C}_{out}, \mathcal{C}_{in})$

$G = (L, R, E)$      $|L| = |R| = n$     d-regular

# AEL Construction

$AEL(G, \mathcal{C}_{out}, \mathcal{C}_{in})$

$G = (L, R, E)$  bipartite  $|L| = |R| = n$   $d$ -regular

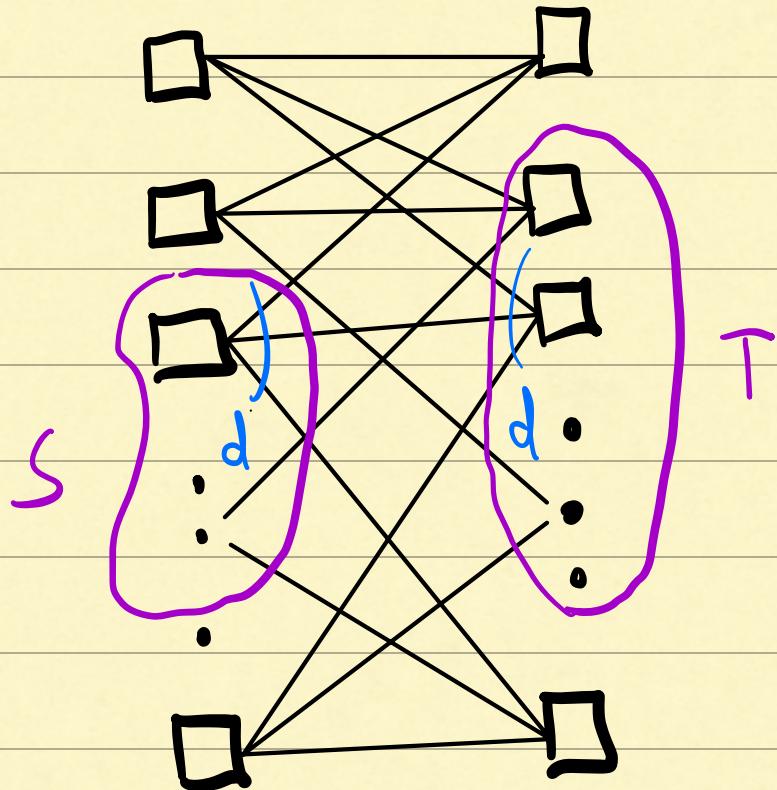


# AEL Construction

$AEL(G, \mathcal{C}_{out}, \mathcal{C}_{in})$

$G = (L, R, E)$  bipartite  $|L| = |R| = n$   $d$ -regular  
 $\lambda$ -expander  $\lambda \in [0, 1]$

$$\left| E(S, T) - \frac{d}{n} |S||T| \right| \leq \lambda d n$$



$\notin$

$S \subseteq L$

$T \subseteq R$

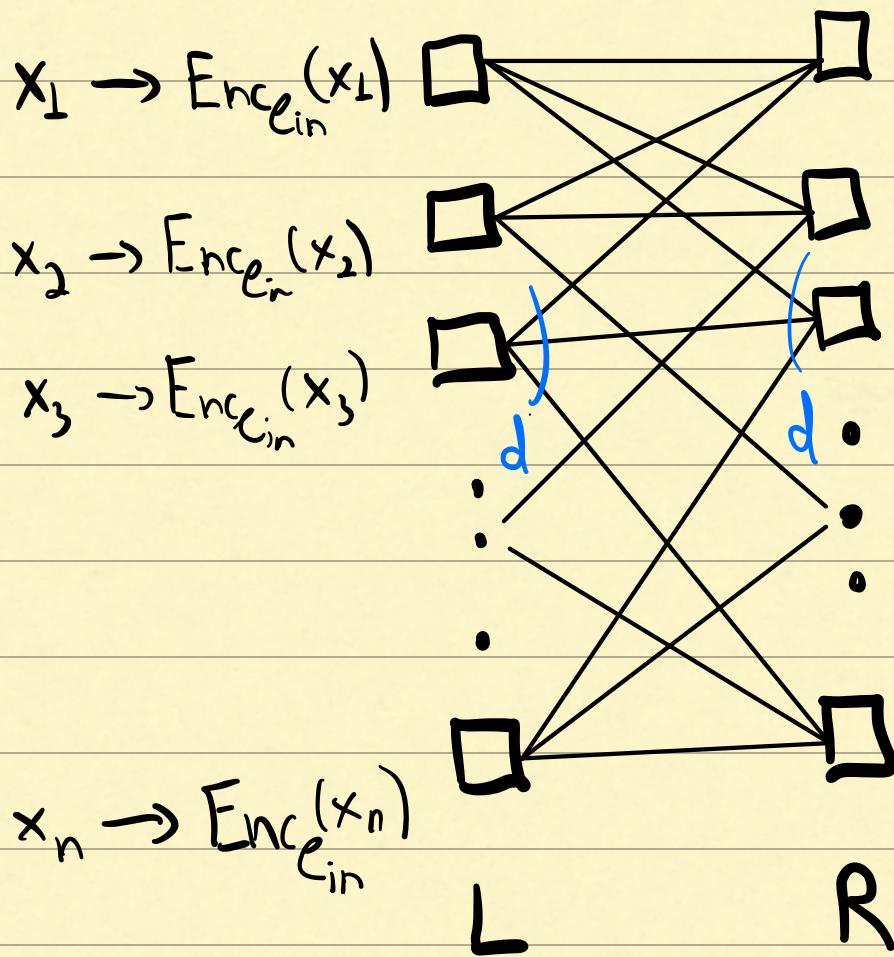
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$G = (L, R, E)$  bipartite  $|L| = |R| = n$   $d$ -regular

$\mathcal{C}_{out} \subseteq \sum_{out}^n, \quad \mathcal{C}_{in} \subseteq \sum_{in}^d$  with  $|\sum_{out}| = |\mathcal{C}_{in}|$   
 $\sum_{out} \cong \mathcal{C}_{in}$

$x = (x_1, \dots, x_n) \in \mathcal{C}_{out}$



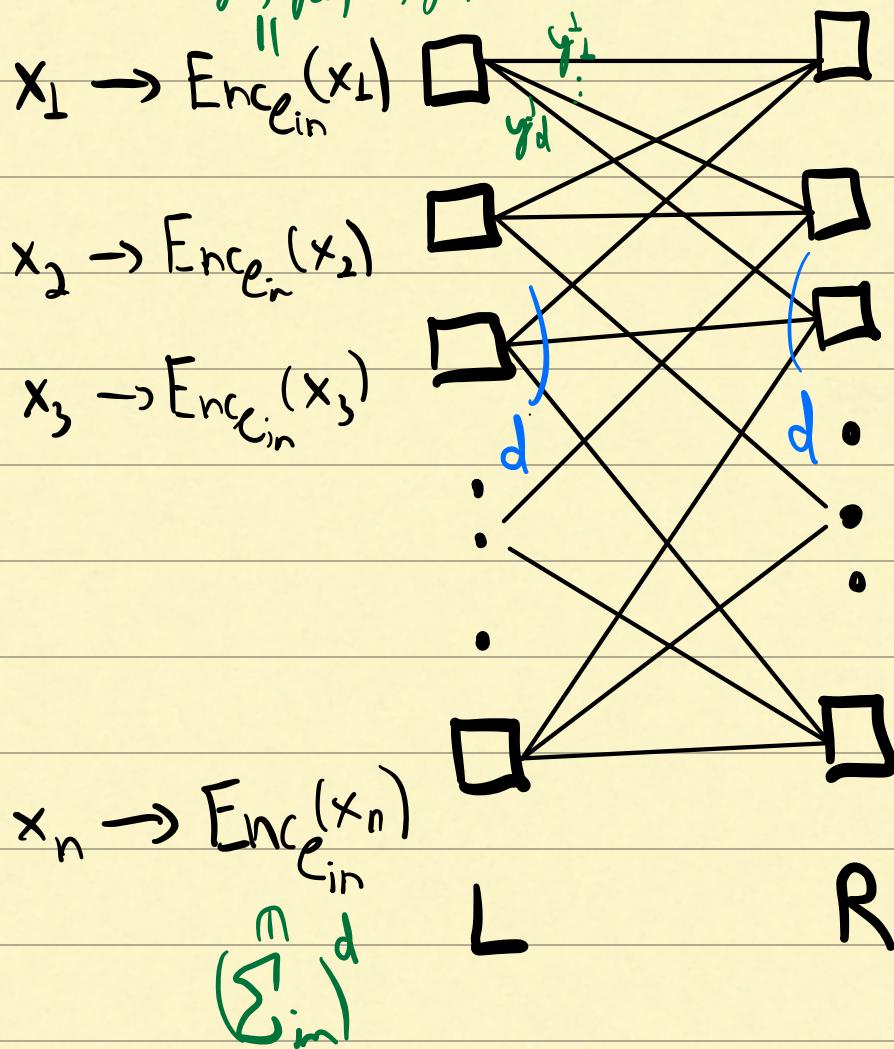
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$x = (x_1, \dots, x_n) \in \mathcal{C}_{\text{out}}$   
 $(y_1, y_2, \dots, y_d) \in (\sum_{\text{in}})^d$



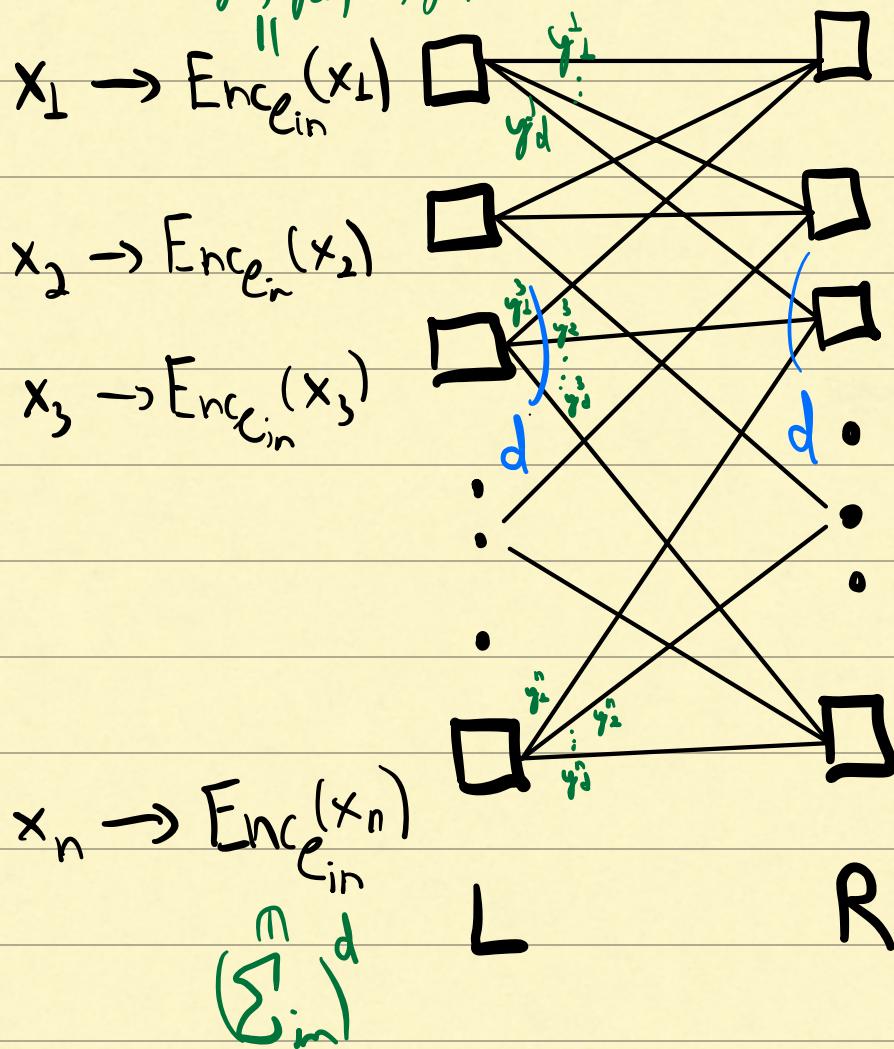
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 $(y_1^1, y_2^1, \dots, y_d^1) \in (\sum_{\text{in}})^d$



# AEL Construction

$\text{AEL}(G, \mathcal{C}_{\text{out}}, \mathcal{C}_{\text{in}})$

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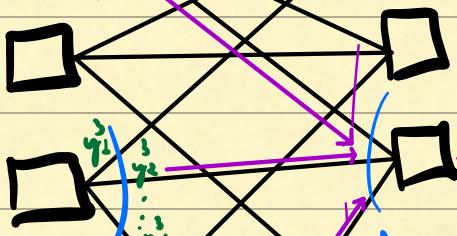
$x = (x_1, \dots, x_n) \in \mathcal{C}_{\text{out}}$

$$(y_1^1, y_2^1, \dots, y_d^1) \in (\sum_{\text{in}})^d$$

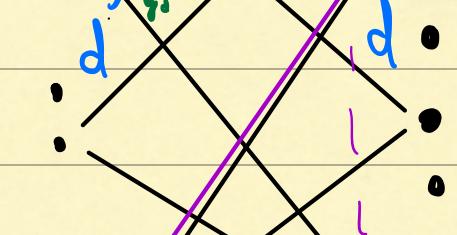
$x_1 \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}(x_1)$



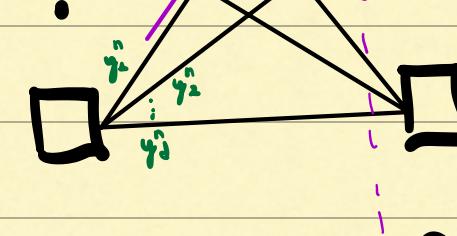
$x_2 \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}(x_2)$



$x_3 \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}(x_3)$



$x_n \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}^{(x_n)}$



$$(\sum_{\text{in}})^d$$

L

R

Collect Symbols

$$(\sum_{\text{in}})^d$$

Collect Symbols

# AEL Construction

$\text{AEL}(G, \mathcal{C}_{\text{out}}, \mathcal{C}_{\text{in}})$

$G = (L, R, E)$  bipartite  $|L| = |R| = n$   $d$ -regular

$\mathcal{C}_{\text{out}} \subseteq \sum_{\text{out}}^n, \quad \mathcal{C}_{\text{in}} \subseteq \sum_{\text{in}}^d$  with  $|\sum_{\text{out}}| = |\mathcal{C}_{\text{out}}| = |\mathcal{C}_{\text{in}}|$   
 $\sum_{\text{out}} \cong \mathcal{C}_{\text{in}}$

$x = (x_1, \dots, x_n) \in \mathcal{C}_{\text{out}}$   
 $(y_1^1, y_2^1, \dots, y_d^1) \in (\sum_{\text{in}})^d$

$x_1 \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}(x_1)$



$x_2 \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}(x_2)$

$x_3 \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}(x_3)$

$x_n \rightarrow \text{Enc}_{\mathcal{C}_{\text{in}}}(x_n)$

$$(\sum_{\text{in}})^d$$

Collect Symbols

$\rightarrow g_1 \in (\sum_{\text{in}})^d$

$\rightarrow g_2$

$\rightarrow g_3 = (y_1^3, y_2^3, \dots, y_d^3)$

$$(\sum_{\text{in}})^d$$

Final codeword

$$(\sum_{\text{in}}^d)^R$$

L

R

How can we analyze the distance?

$$g, g' \in \mathcal{C}_{AEL} \subseteq (\Sigma_{in}^d)^R$$

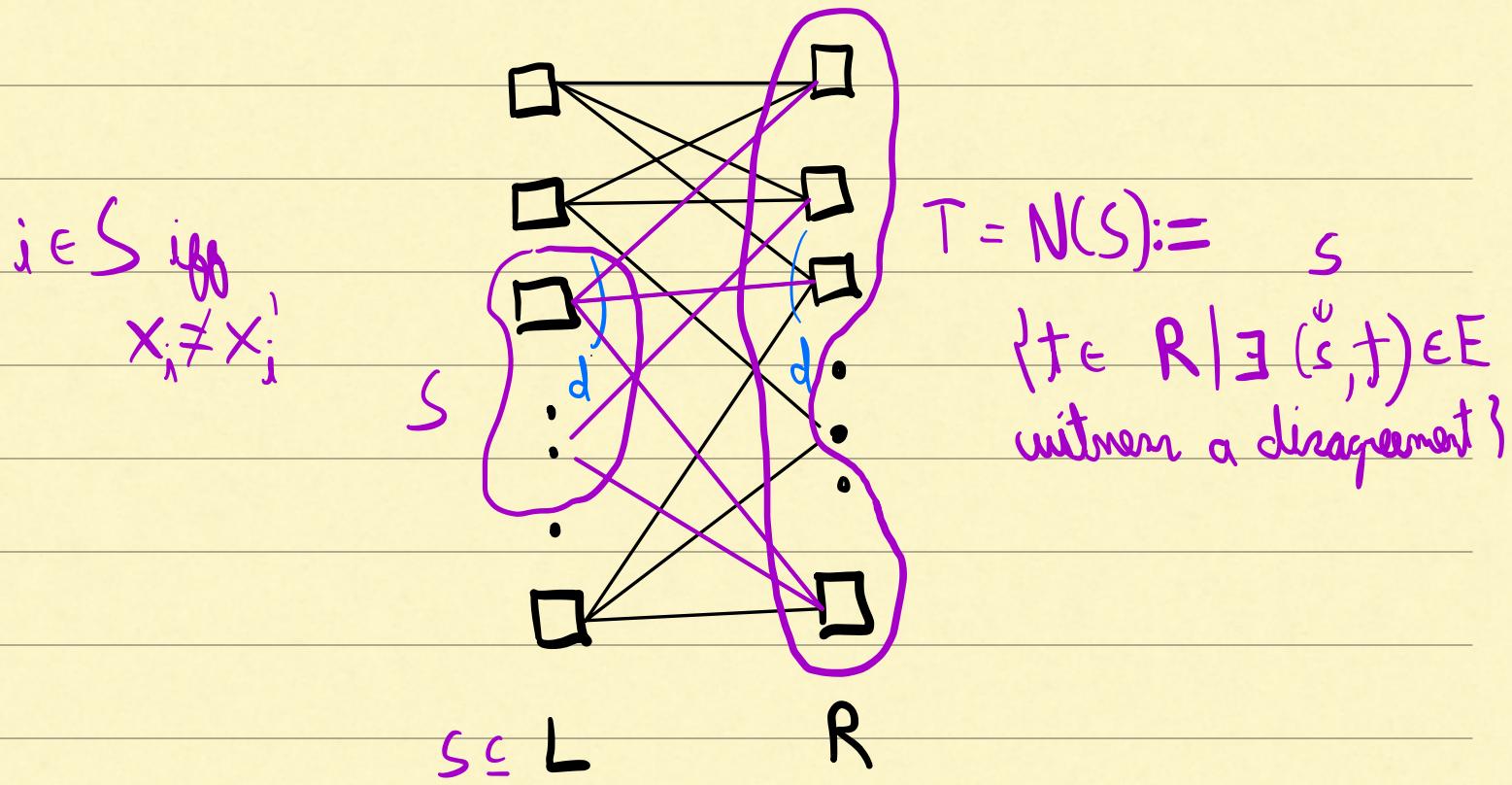
$$g \neq g'$$

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$\uparrow$        $\uparrow$   
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$$x, x' \in \mathcal{C}_{out} \quad \Delta(x, x') \geq \delta(\mathcal{C}_{out})$$



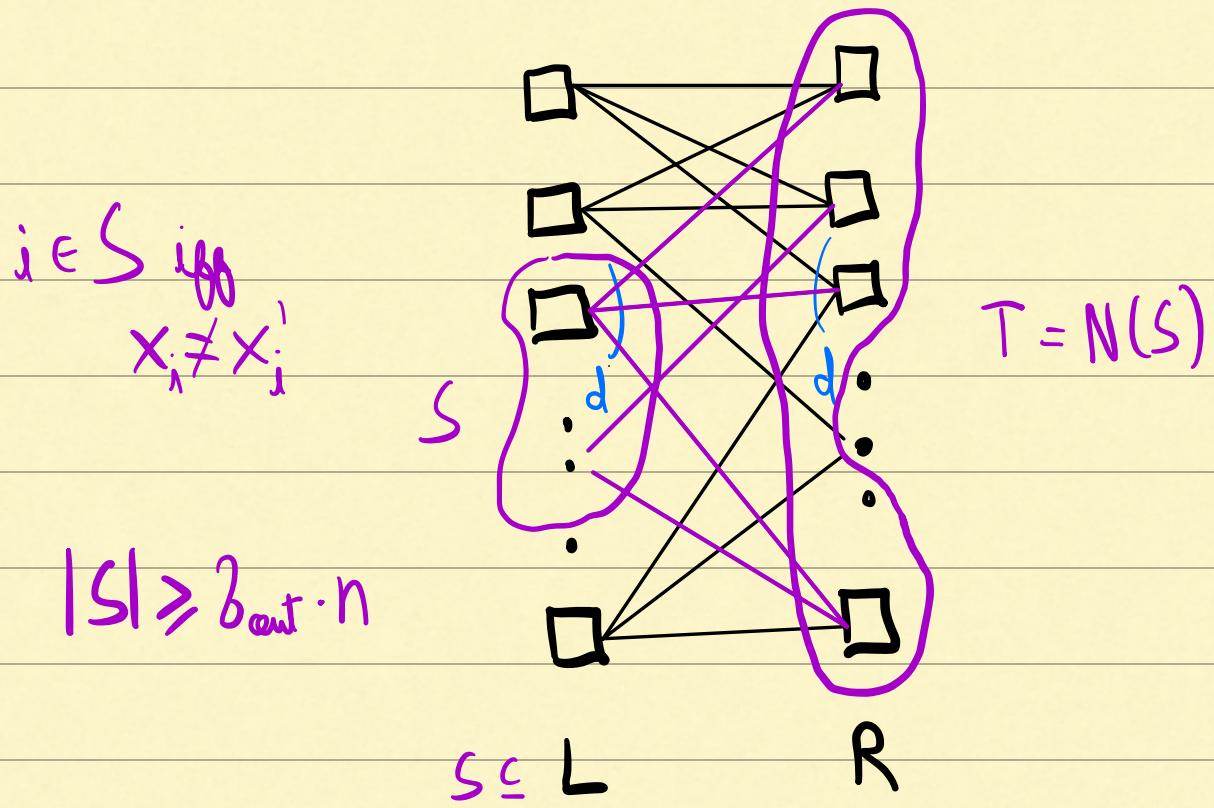
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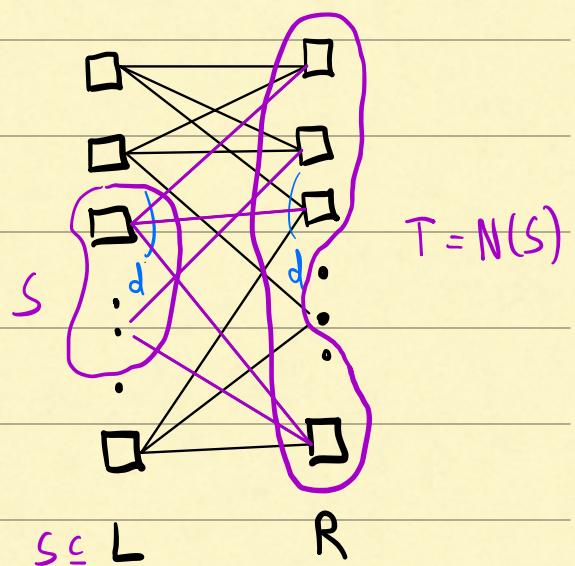
$$x, x' \in \mathcal{C}_{out}$$

$$\Delta(x, x') \geq \delta_{out}$$

$$\Delta(g, g') \geq |T|/n$$

$$i \in S \text{ iff } x_i \neq x'_i$$

$$|S| \geq \delta_{out} \cdot n$$



$$S \subseteq L$$

$$R$$

$$|E(S, T) - \frac{d}{n} |S||T|| \leq \lambda d n$$

$$E(S, T) \leq \frac{d}{n} |S||T| + \lambda d n$$

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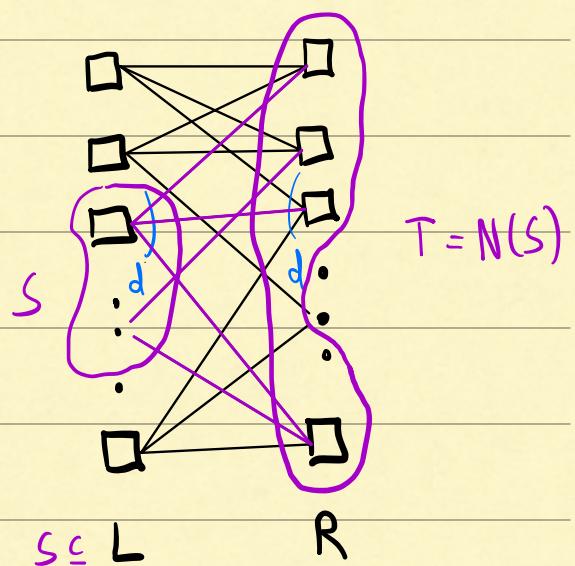
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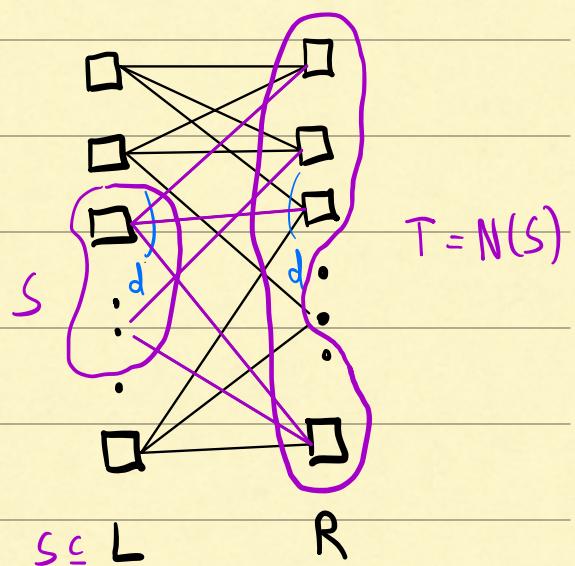
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$$\underline{\delta_{in}} - \lambda n \frac{|S|}{|S|} \leq \frac{|T|}{n}$$

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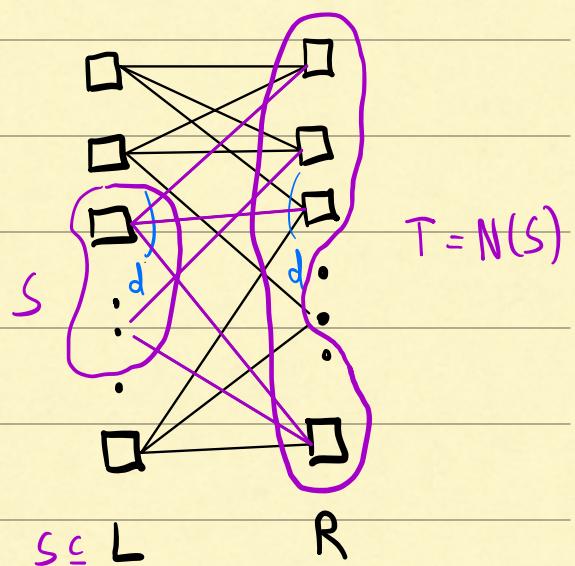
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$$|E(S, T)| - \frac{d}{n} |S||T| \leq \lambda d n$$

$$\delta_{in} d |S| \leq E(S, T) \leq \frac{d}{n} |S||T| + \lambda d n$$



$$\delta_{in} - \lambda n \frac{|S|}{|T|} \leq \frac{|T|}{n}$$

$$|S| \geq \delta_{out} \cdot n$$



$$\delta_{in} - \lambda \frac{\delta_{out}}{n} \leq \frac{|T|}{n} \leq \Delta(g, g')$$

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$$g, g' \in \mathcal{C}_{AEL} \subseteq (\Sigma_n^d)^R$$

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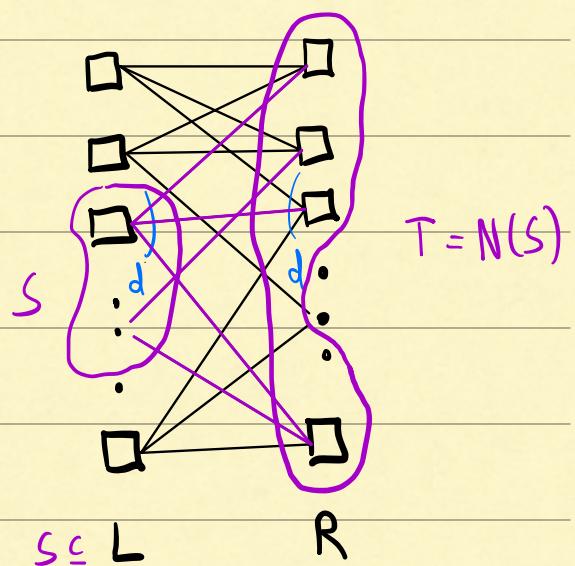
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$$\delta_{in} d |S| \leq E(S, T) \leq \frac{d}{n} |S||T| + \lambda d n$$



$$\delta_{in} - \frac{\lambda n}{|S|} \leq \frac{|T|}{n}$$

$$|S| \geq \delta_{out} \cdot n$$

$$\delta_{in} - \frac{\lambda}{\delta_{out}} \approx 0 \leq \frac{|T|}{n} \leq \Delta(g, g')$$

# AEL Parameters

$$\beta(e_{AEL}) \geq \beta_{in} - \cancel{\beta_{out}} \approx 0$$

$$r(e_{AEL}) = r(e_{out}) r(e_{in})$$

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We can take  $e_{in}$  with  $\beta_{in} \geq 1-R$

$$r(e_{in}) \geq R$$

and  $e_{out}$  with  $r(e_{out}) \geq 1-\epsilon$

# AEL Parameters

$\lambda \rightarrow 0$

$$\delta(e_{AEL}) \geq \delta_{in} - \cancel{\delta_{out}} \approx 0$$

$$r(e_{AEL}) = r(e_{out}) r(e_{in})$$

We can take  $e_{in}$  with  $\delta_{in} \geq 1-R$

$$r(e_{in}) \geq R$$

and  $e_{out}$  with  $r(e_{out}) \geq 1-\epsilon$

$$\delta(e_{AEL}) \geq 1-R-\epsilon$$

$$r(e_{AEL}) \geq R-\epsilon$$

near the Singleton bound

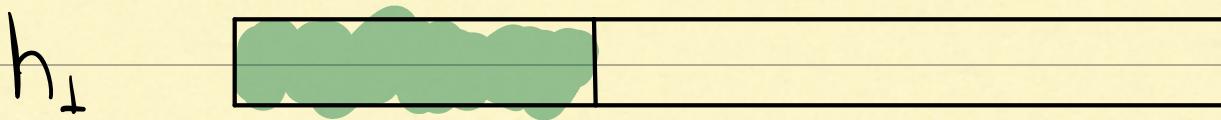
Is there a generalization of  
the Singleton bound for  
list decoding?

# Generalized Singleton Bound

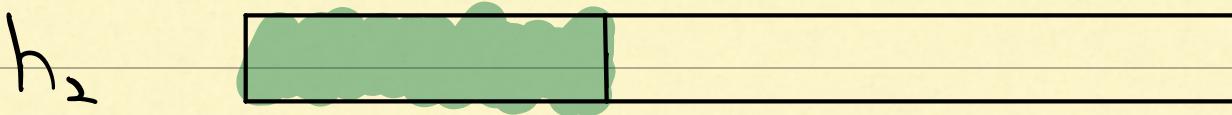
Let  $C \subseteq \Sigma^n$  of rate  $R \in [0,1]$   
 $(|C| = |\Sigma|^{Rn})$

$h_1, h_2, h_3 \in C$  (distinct codewords)

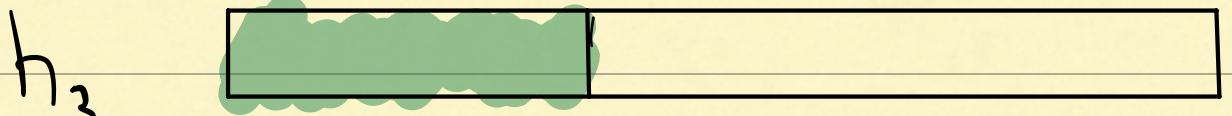
$R_{n-1}$  :



|| |



|| |

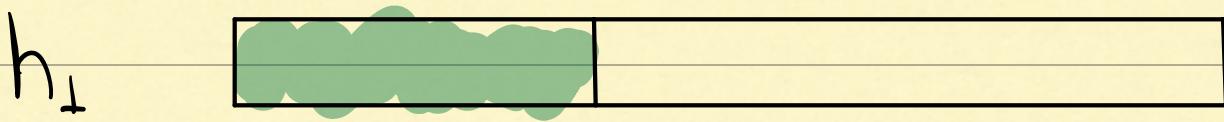


## Generalized Singleton Bound

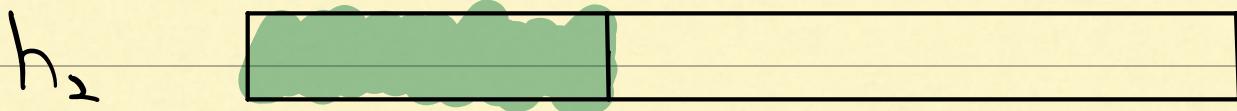
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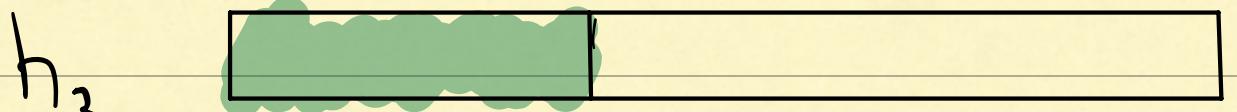
$R_{n-1}$  :



|| |



|| |



Can we construct a center  $g \in \Sigma^n$

close to  $h_1, h_2, h_3$ ?

# Generalized Singleton Bound

Let  $C \subseteq \Sigma^n$  of rate  $R \in [0,1]$   
 $(|C| = |\Sigma|^{Rn})$

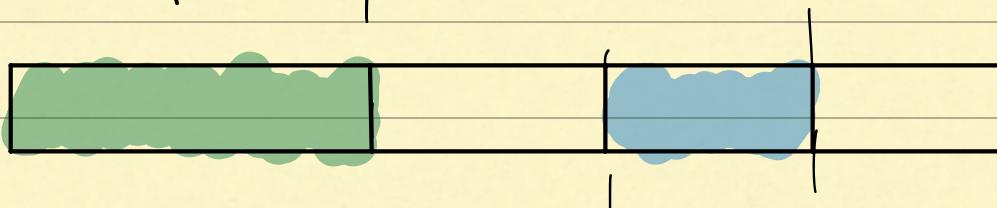
$h_1, h_2, h_3 \in C$  (distinct codewords)

$$Rn-1 \quad \underbrace{\quad\quad\quad}_{(1-R)n+1}$$

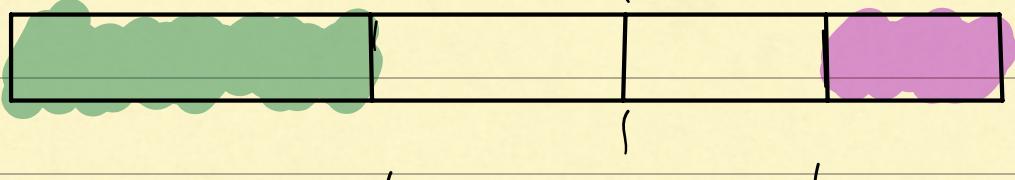
$h_1$



$h_2$



$h_3$



$g$



$h_1, h_2, h_3 \in \mathcal{L}(g, \frac{2}{3} (1-R + \frac{1}{n}))$

# Generalized Singleton Bound

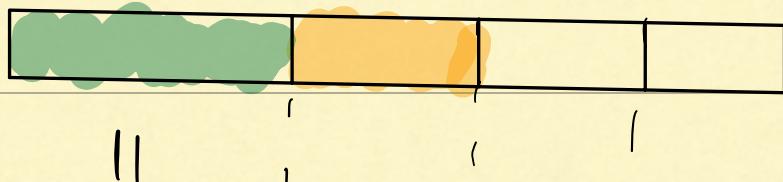
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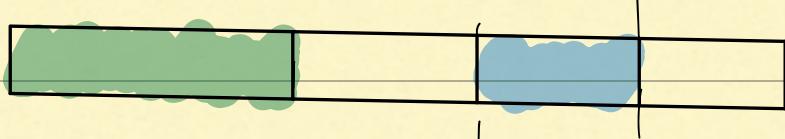
$$(1-R)n+1$$

$Rn-1$

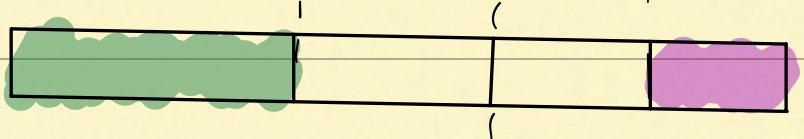
$h_1$



$h_2$



$h_3$



$g$



$$\left| \text{d}(g, \frac{2}{3} (1-R)) \right| > 3$$

# Generalized Singleton Bound

Let  $C \subseteq \Sigma^n$  of rate  $R \in [0,1]$   
 $(|C| = |\Sigma|^{Rn})$

$h_1, h_2, h_3 \in C$  (distinct codewords)

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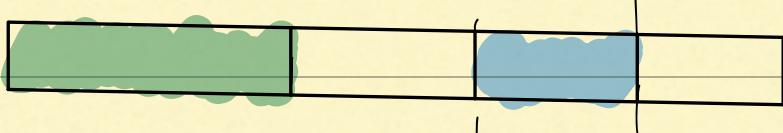
||

||

||

||

$h_2$



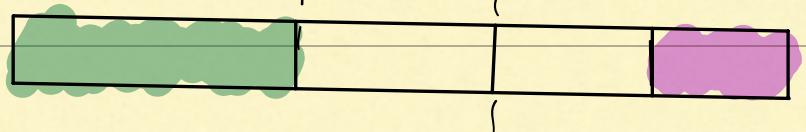
||

||

||

||

$h_3$



||

||

||

||

$g$



Similar argument holds for K codewords

$$\left| \mathcal{I}(g, \underbrace{(K-1)(1-R)}_{K}) \right| \geq K$$

# Generalized Singleton Bound (GSB)

Thm [Shengguan-Tamo'20, Roth'22]

$$\left| \mathcal{I}\left(g, \frac{(k-1)(1-R)}{K} \right) \right| \geq k$$

---

$K=2$  in the original Singleton Bound

---

# Generalized Singleton Bound (GSB)

Thm [Shengguan-Tamo'20, Roth'22]

$$\left| \mathcal{L}\left(g, \frac{(k-1)(1-R)}{K}\right) \right| \geq K$$

---

$K=2$  in the original Singleton Bound

---

$$\text{Con: } \left| \mathcal{L}\left(g, 1-R-\varepsilon\right) \right| \geq \Omega\left(\frac{1}{\varepsilon}\right)$$

---

Large  $\Omega\left(\frac{1}{\varepsilon}\right)$  are required  
for capacity

---

Moral:

Achieving list size of  $O\left(\frac{1}{\varepsilon}\right)$  would  
be optimal

# Folded Reed-Solomon (FRS)

List size bounds

-  $n^{\frac{1}{\varepsilon}}$  [Guruswami - Rudra '06]  
[Guruswami '11]

-  $\left(\frac{1}{\varepsilon}\right)^{\frac{1}{\varepsilon}}$  [Bri - Lovett '12]

-  $\frac{1}{\varepsilon^2}$  [Srivastava '25]

-  $\frac{1}{\varepsilon}$  [Chen - Zhang '25]  
(approach GSB)

# Folded Reed-Solomon (FRS)

## Wonderful Properties

- List Decoding Capacity ✓  
[Guruswami - Rudra '06]
- Explicit ✓
- Efficient (List) Decoding ✓
- Approach GSB  $\Rightarrow$  optimal list sizes ✓  
[Chen - Zhang '25]

## Major Drawbacks

- Large non-constant alphabet size  
 $(C \subseteq \mathbb{F}_q^n \text{ with } q = n^{\frac{1}{\varepsilon^2}}) \wedge$
- No good "local" properties (LDPC) ↴

# Daymn of Papers Studying

## List Decoding (Capacity) for

### Random Codes

- Powerful yardstick to analyze ✓ constructions
- Beautiful Combinatorial/Probabilistic ✓ Ideas
  - :
  - :

# Dozens of Papers Studying

## List Decoding (Capacity) for

### Random Codes

- Powerful yardstick to analyze ✓ constructions
- Beautiful Combinatorial/Probabilistic ✓ Ideas  
:  
:
- Cannot efficiently certify  $\Delta(e)$  ↴
- Cannot efficiently decode ↴
- Cannot be made explicit ↴
- Cannot use in real applications ↴

# Dream Wish List of Coding Theory

We want a code with:



- Optimal Parameters  
(Rate-vs-decoding radius)
- Explicit
- Alphabet as small as possible
- Efficient Encoding
- Efficient (list) Decoding
- Local Properties  
(LDPC, LTC, LDC, LCC, etc)

## Our Results

Thm [J-Mittal-Srivastava-Tuliani'25]

New family of codes

- Explicit

- Achieve List Decoding Capacity

- Over constant size alphabet!

- Optimal List Sizes (approach GSB)!

- "Efficient" List Decoding up to Capacity  
(Sum-of-Squares based)

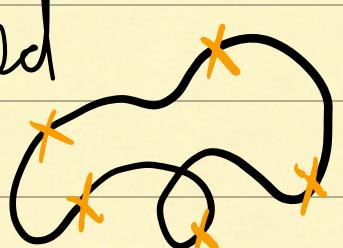
- Local Properties: LDPC!

# Known Codes satisfying

- Explicit ✓ [Guo-Rom-Zewi '20]
- Achieve List Decoding Capacity ✓
- Over constant size alphabet ✓

are based on sophisticated

Algebraic Geometry



and still have the drawbacks:

- List sizes are far from optimal ↗
- Do not approach GSB ↗

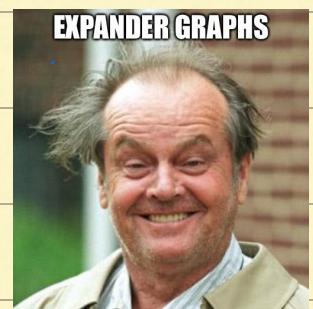
## Our Results

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New family of codes

- Explicit
- Achieve List Decoding Capacity
- Over constant size alphabet!
- Optimal List Sizes (approach GSB)!
- "Efficient" List Decoding up to Capacity  
(Sum-of-Squares based)
- Local Properties: LDPC!

Based on Expander Graphs  
(AEL for list decoding)



- Simple Construction
- Only uses Combinatorics  
(No need of sophisticated algebra)

## Our Codes

Local-to-Global (using AEL)

Start with a constant ring  $\mathcal{C}_n$  approaching  
the generalized Singleton Bound\* (GRS)



obtain via AEL

Global Code approaching the GRS

# Our Codes

## Local-to-Global (using AEL)

Start with a constant ring  $\mathcal{C}_{\text{in}}$  approaching the generalized Singleton Bound\* (GRS)



obtain via AEL

Global Code approaching the GRS

---

(1)  $\mathcal{C}_{\text{in}}$  can be a random linear code of  $O(1)$  size

or

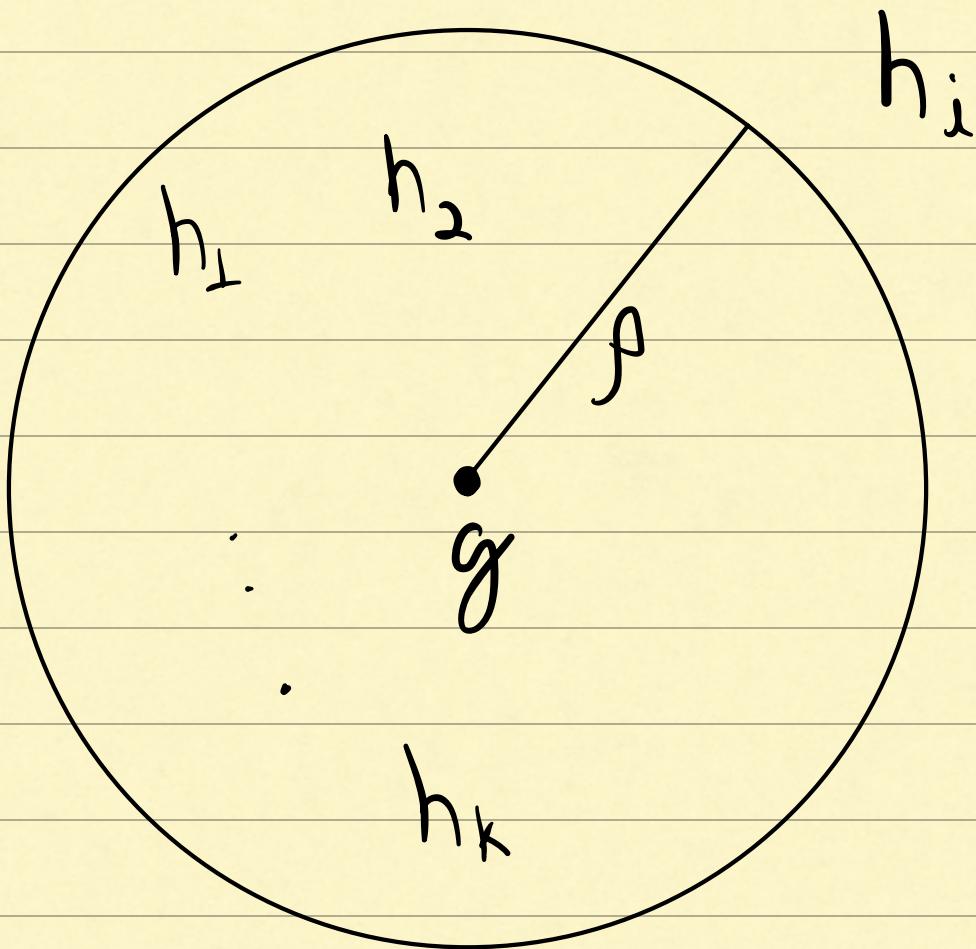
(2)  $\mathcal{C}_{\text{in}}$  can be an explicit FRS of  $O(1)$  size

## List Decoding

Def:  $C \subseteq \Sigma^n$  is  $(p, k-1)$ -list decodable if

$\forall g \in \Sigma^n, h_1, \dots, h_k \in C$

$$\max_{i \in [k]} \Delta(g, h_i) > p$$



"Not all  $h_1, \dots, h_k$  can be in  $B(g, p)$ "

# Average Radius List Decoding (stronger than list decoding)

Def:  $C \subseteq \Sigma^n$  is  $(p, k-1)$ -list decodable if

$\forall g \in \Sigma^n, h_1, \dots, h_k \in C$  (distinct)

$$\left| \bigcap_{i \in [k]} \Delta(g, h_i) \right| > p$$

# Average Radius List Decoding (stronger than list decoding)

Def:  $C \subseteq \Sigma^n$  is  $(p, k-1)$ -list decodable if

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$$\left| \sum_{i \in [k]} \Delta(g, h_i) \right| > p$$

## $\epsilon$ -relaxed average GSB

$$\left| \sum_{i \in [k]} \Delta(g, h_i) \right| \geq \frac{(k-1)}{k} (1-R-\epsilon)$$



$$\sum_{i \in [k]} \Delta(g, h_i) \geq (k-1)(1-R-\epsilon)$$

# Glimpse of The Proof

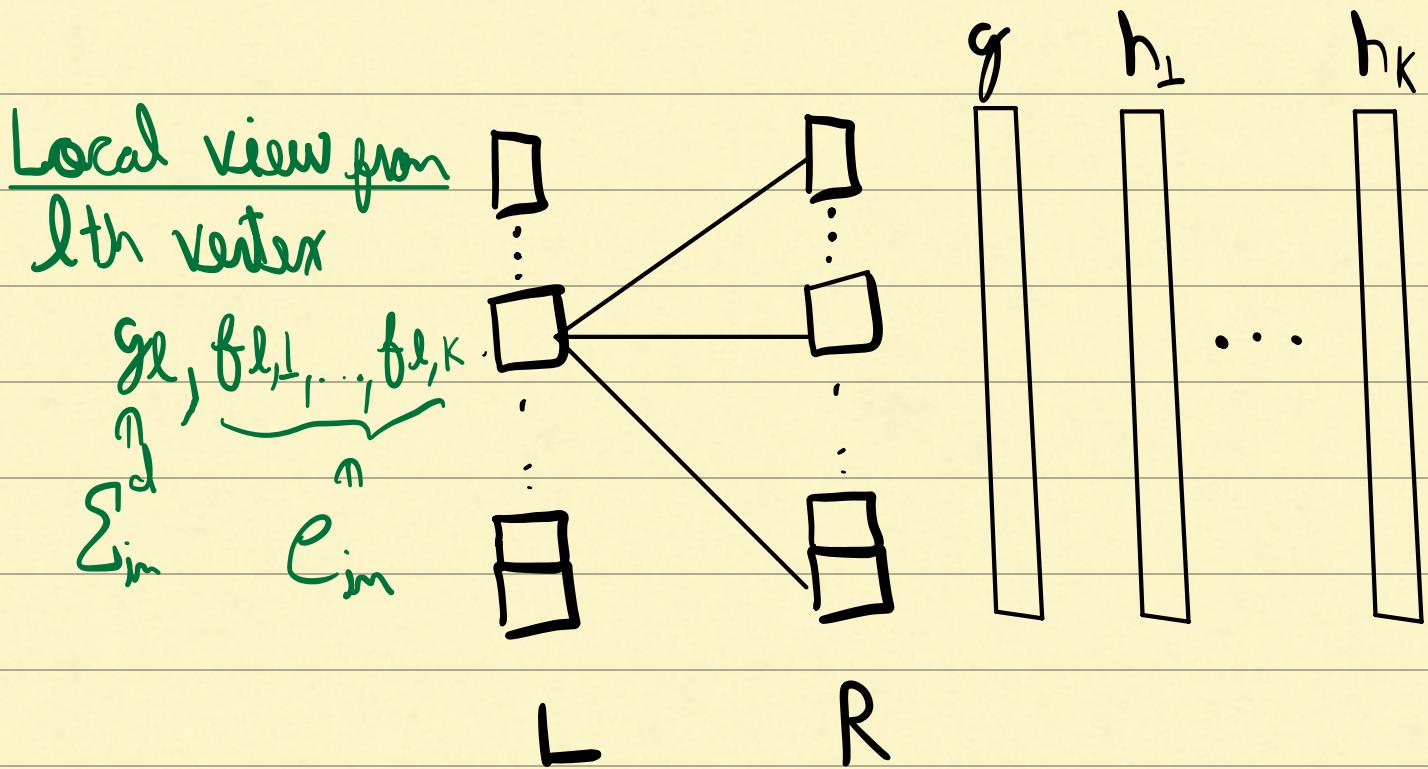
$AEL(g, \mathcal{L}_{out}, \mathcal{L}_{in})$

$\vdash \varepsilon\text{-relaxed 6SB}$

$$\nexists f \in \sum_{in}^d, f_1, \dots, f_k \in \mathcal{L}_{in}, \sum_{i \in [k]} \Delta(f, f_i) \geq (k-1)(1-R-\varepsilon)$$

$$g \in (\sum_{in}^d)^r, h_1, \dots, h_k \in \mathcal{L}_{AEL}$$

distinct



# Glimpse of The Proof

$AEL(g, \mathcal{C}_{out}, \mathcal{C}_{in})$

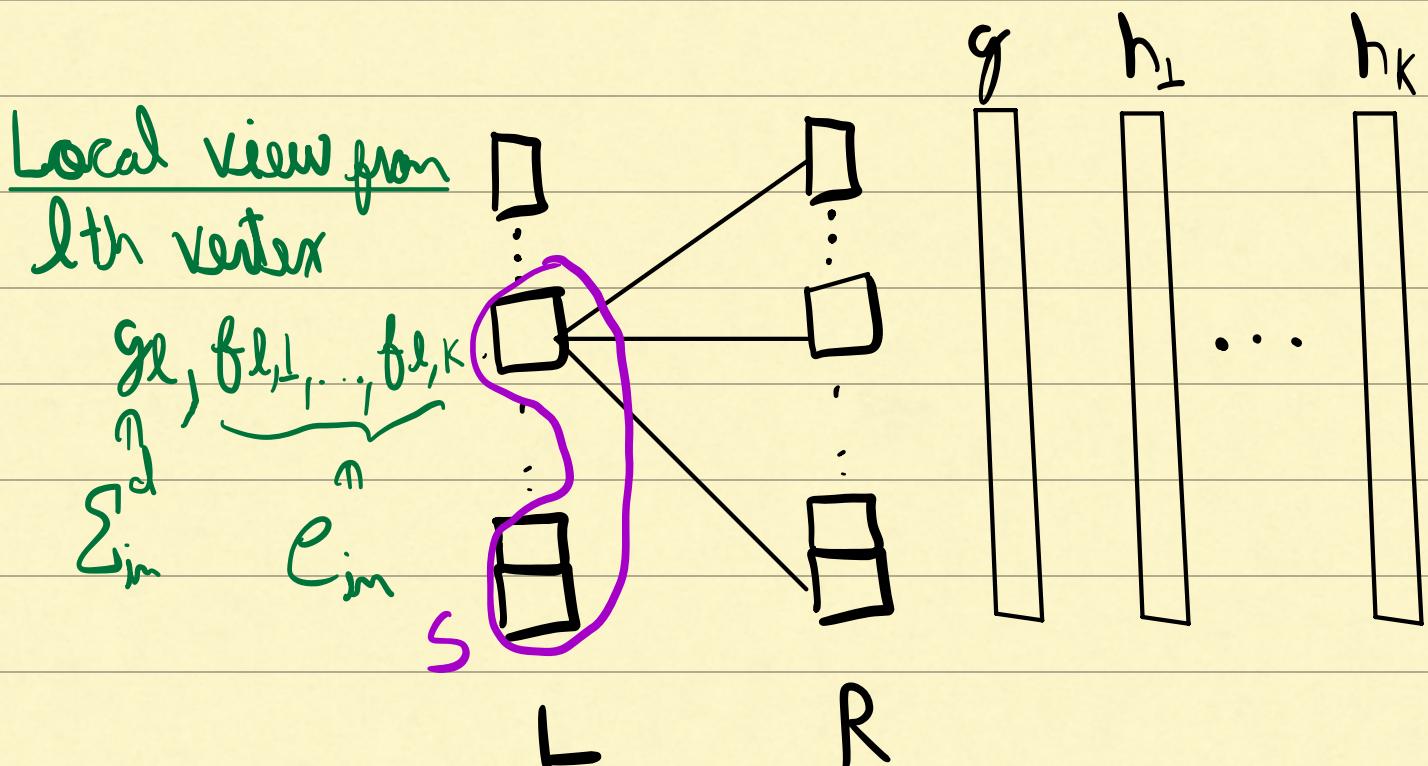
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distinct

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distinct



$S = \{l \in [n] \mid f_{l,1}, \dots, f_{l,k} \text{ are not all equal}\}$

# Glimpse of The Proof

$AEL(g, \mathcal{C}_{out}, \mathcal{C}_{in})$

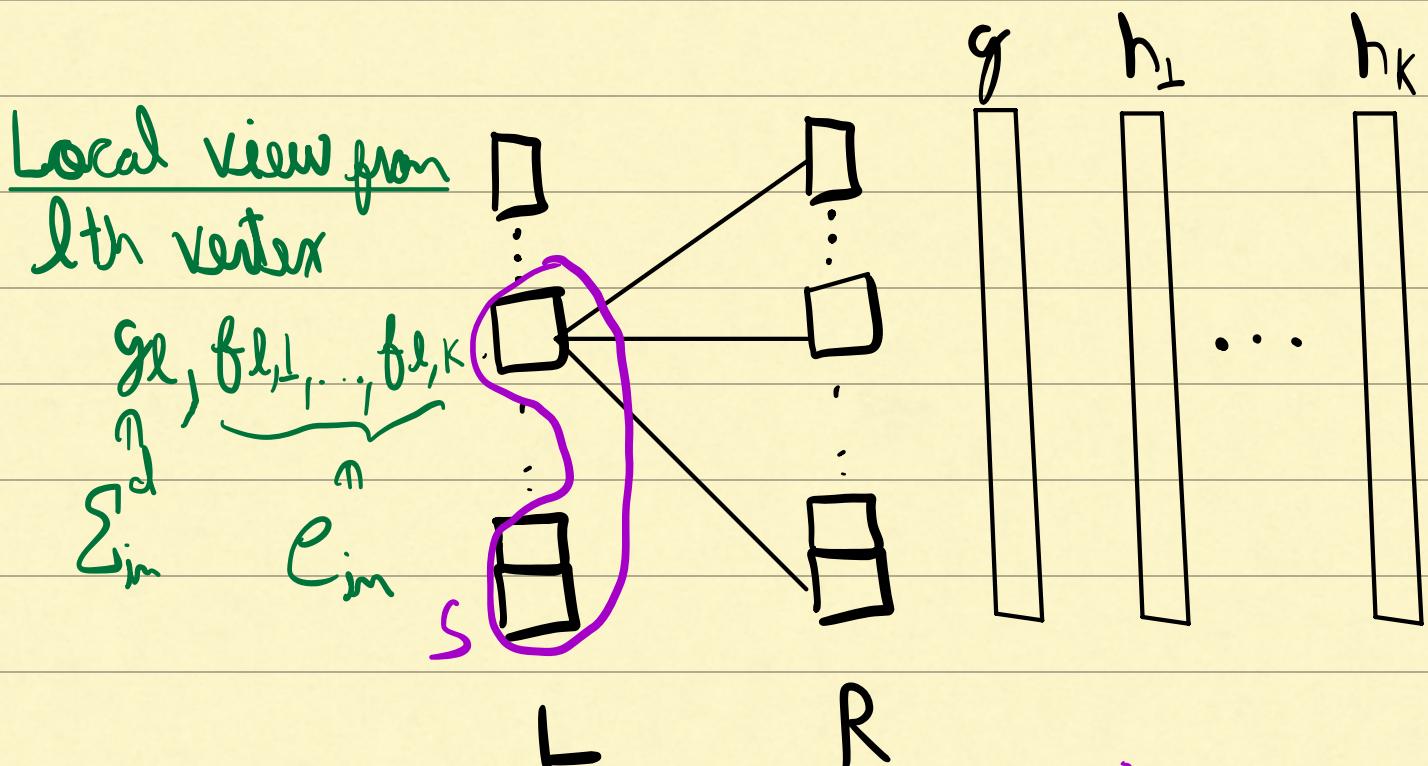
$\vdash \varepsilon\text{-relaxed } GS$

$$\nexists f \in \sum_{in}^d, f_1, \dots, f_k \in \mathcal{C}_{in}, \sum_{i \in [k]} \Delta(f, f_i) \geq (k-1)(1-R-\varepsilon)$$

distinct

$$g \in (\sum_{in}^d)^n, h_1, \dots, h_k \in \mathcal{C}_{AEL}$$

distinct



$$S = \{l \in [n] \mid f_{l,1}, \dots, f_{l,k} \text{ are not all equal}\}$$

$$|S| \geq \beta_{out} \cdot n$$

# Glimpse of The Proof

$AEL(g, \mathcal{C}_{out}, \mathcal{C}_{in})$

$\vdash \varepsilon\text{-relaxed } 6SB$

$\nexists f \in \sum_{in}^d, f_1, \dots, f_k \in \mathcal{C}_{in}, \sum_{i \in [k]} \Delta(f, f_i) \geq (k-1)(1-R-\varepsilon)$

$g \in (\sum_{in}^d)^n, h_1, \dots, h_k \in \mathcal{C}_{AEL}$

distinct

Local view from  
l<sup>th</sup> vertex

$g_l, f_{l,1}, \dots, f_{l,k}$

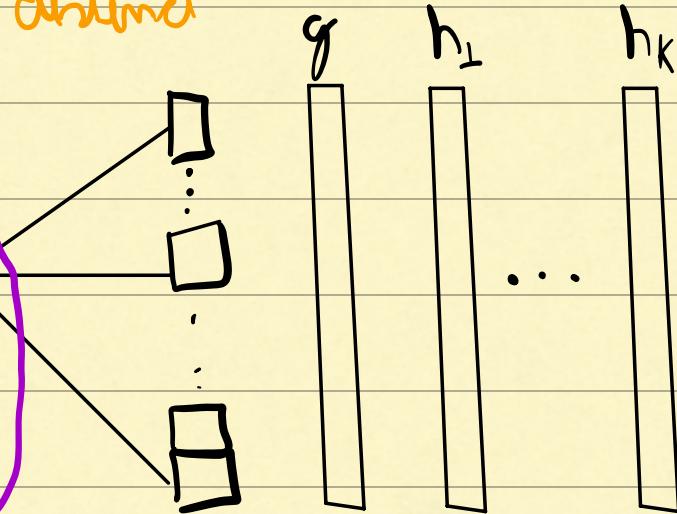
$\sum_{in}^d$

$\mathcal{C}_{in}$

S

L

R



$S = \{l \in [n] \mid f_{l,1}, \dots, f_{l,k} \text{ are not all equal}\}$

$|S| \geq 3_{out} \cdot n$

Easy Case:  $\nexists l \in S, |f_{l,1}, \dots, f_{l,k}| = k$

# Glimpse of The Proof

$AEL(g, \mathcal{C}_{out}, \mathcal{C}_{in})$

$\vdash \varepsilon\text{-relaxed } 6SB$

$$\nexists f \in \sum_{in}^d, f_1, \dots, f_k \in \mathcal{C}_{in}, \sum_{i \in [k]} \Delta(f, f_i) \geq (k-1)(1-R-\varepsilon)$$

$$g \in (\sum_{in}^d)^n, h_1, \dots, h_k \in \mathcal{C}_{AEL}$$

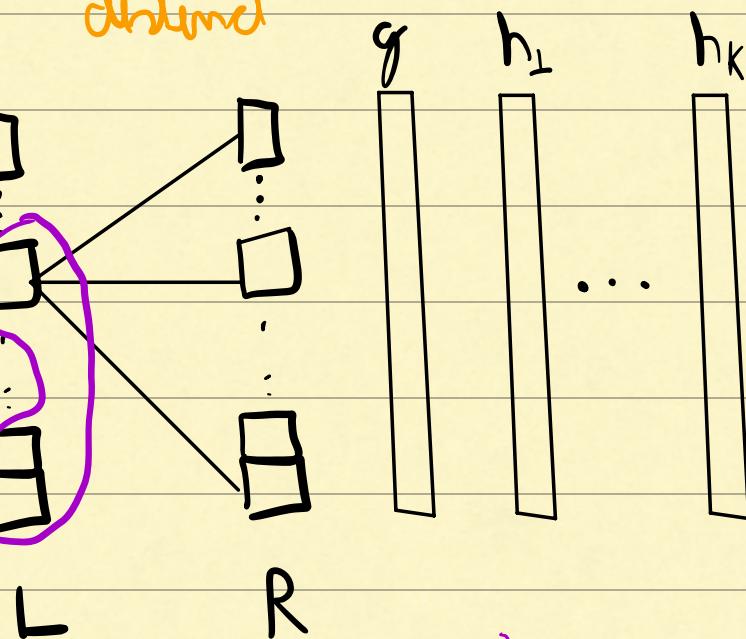
distinct

Local view from  
l<sup>th</sup> vertex

$$g_l, f_{l,1}, \dots, f_{l,k}$$

$\sum_{in}^d$

$\mathcal{C}_{in}$



$$S = \{l \in [n] \mid f_{l,1}, \dots, f_{l,k} \text{ are not all equal}\} \quad |S| \geq \delta_{out} \cdot n$$

Easy Case:  $\nexists l \in S, |\{f_{l,1}, \dots, f_{l,k}\}| = k$

$$\Delta(g, h_j) \geq \sum_{l \in S} \Delta(g_l, f_{l,j}) \quad \forall j \in [k]$$

# Glimpse of The Proof

AEL( $g, \mathcal{C}_{\text{out}}, \mathcal{C}_{\text{in}}$ )

$\vdash \varepsilon\text{-relaxed 6SB}$

$$\nexists f \in \sum_{\text{in}}^d, f_1, \dots, f_k \in \mathcal{C}_{\text{in}}, \sum_{i \in [k]} \Delta(f, f_i) \geq (k-1)(1-R-\varepsilon)$$

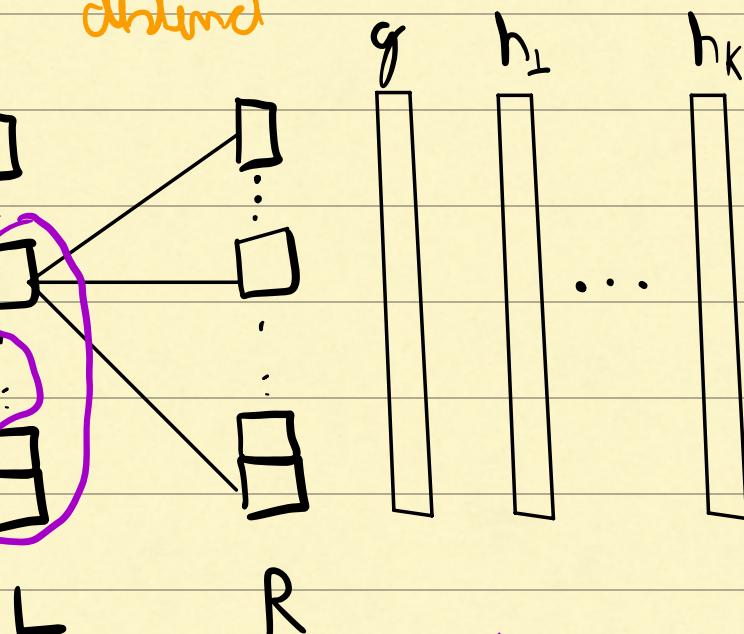
distinct

$$g \in (\sum_{\text{in}}^d)^n, h_1, \dots, h_k \in \mathcal{C}_{\text{AEL}}$$

distinct

Local view from  
l<sup>th</sup> vertex

$$g_l, f_{l,1}, \dots, f_{l,k} \\ \sum_{\text{in}}^d \\ \mathcal{C}_{\text{in}}$$



$$S = \{l \in [n] \mid f_{l,1}, \dots, f_{l,k} \text{ are not all equal}\}$$

$$|S| \geq \delta_{\text{out}} \cdot n$$

Easy Case:  $\nexists l \in S, |f_{l,1}, \dots, f_{l,k}| = k$

$$\sum_{i \in [k]} \Delta(g, h_i) \geq |E| \sum_{l \in S} \sum_{i \in [k]} \Delta(g_l, f_{l,i})$$

# Glimpse of The Proof

AEL( $g, \mathcal{C}_{\text{out}}, \mathcal{C}_{\text{in}}$ )

$\vdash \varepsilon$ -relaxed 6SB

$$\nexists f \in \sum_{\text{in}}^d, f_1, \dots, f_k \in \mathcal{C}_{\text{in}}, \sum_{i \in [k]} \Delta(f, f_i) \geq (k-1)(1-R-\varepsilon)$$

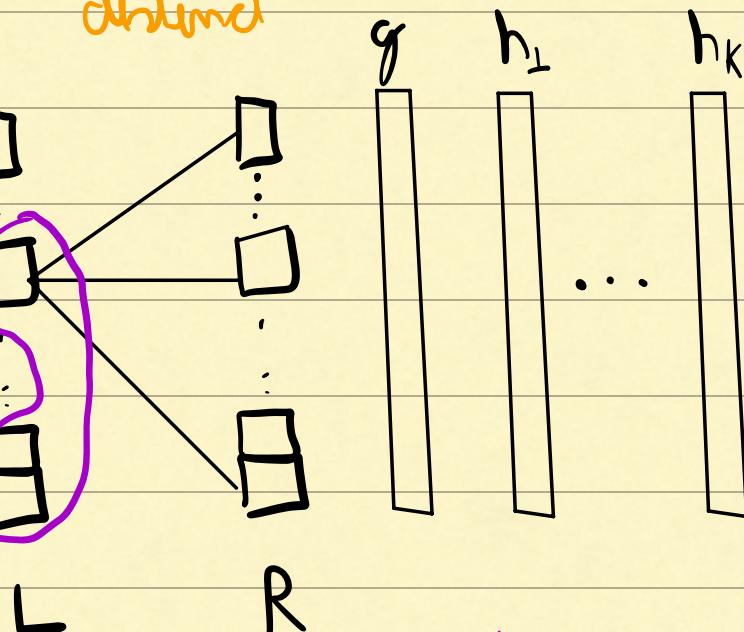
$$g \in (\sum_{\text{in}}^d)^n, h_1, \dots, h_k \in \mathcal{C}_{\text{AEL}}$$

distinct

Local view from  
lth vertex

$$g_l, f_{l,1}, \dots, f_{l,k}$$

$\sum_{\text{in}}^d$



$$S = \{l \in [n] \mid f_{l,1}, \dots, f_{l,k} \text{ are not all equal}\}$$

$$|S| \geq \delta_{\text{out}} \cdot n$$

Easy Case:  $\nexists l \in S, |f_{l,1}, \dots, f_{l,k}| = k$

$$\sum_{i \in [k]} \Delta(g, h_i) \geq (k-1)(1-R-\varepsilon)$$



## Open Question Galore

- More efficient Decoding
- Implications to other pseudorandom constructions  
(e.g., extractors, lossless expanders)
- Tighter alphabet size
- Quantum Capacity
- List Recovery / Other models

# Open Problems

Thank you!

