

# REED MULLER CODES ACHIEVE CAPACITY ON ERASURE CHANNELS

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*E2:207:Concentration Inequalities*

*Course Project Report*

*by*

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# Motivation

The channel capacity is the largest rate at which communication *can be* made with arbitrary small error.

Channel coding tells us **how** capacity can be achieved.

Turbo , SC-LDPC achieve capacity, but no proof is presented yet

Polar and Reed-Muller were proved to achieve Capacity...

- Another\* answer to the search for PROVABLY capacity achieving channel codes.
- Perhaps one of the most important applications of *threshold phenomenon*.
- Establishes that a certain class of codes achieve capacity, Which include Reed-Muller codes, Polar codes.

*\*That polar codes achieve capacity was proved by E.Arikan before this.*

*This provides an alternate method of the same proof leveraging fewer properties of polar codes*

# Channel Capacity from the perspective of Threshold Phenomenon...*proof idea*

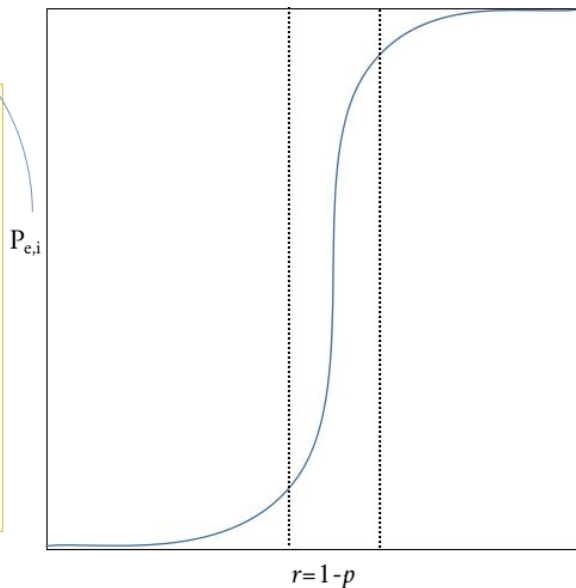
For transmission over BEC(p), with Capacity  $r=1-p$

For a sequence of binary linear codes with rate  $r$  to be capacity achieving, the bit error probability, must converge to 0 for any erasure rate below  $1 - r$ .

if ,  
1.  $P_{e,i} = h(p) = P_p(\Omega_i)$

2. where  $\Omega_i$  is a symmetric monotone , we can apply Russo's lemma.

3. If the critical value is  $p=1-r$ , the proof is complete



*To prove this using Threshold Phenomenon.*

1. *Bit Error Probability* for a code , under MAP decoding needs to be captured by a function of erasure probability. We shall explore *MAP EXIT (Extrinsic Information Transfer) functions* for this.
2. MAP EXIT function should be expressed as measure of a *monotone set*
3. The *influences of the set must be equal* , in other words ,the set must be *symmetric*. This happens when we assume the *code is 2-transitive* under automorphic permutation.
4. Hence we will apply *Russo's lemma*.
5. The critical value will occur at  $p=1-r$ , this follows, from *Area theorem*
6. RM codes , Polar codes are 2-transitive.

# Bit Map Decoding under BEC

## *Decoding rule*

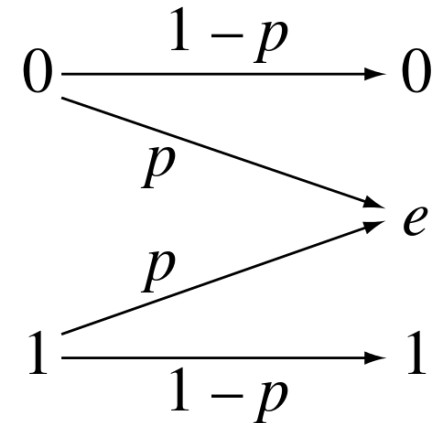
$$D_i : \mathcal{Y}^N \rightarrow \mathcal{X} \cup \{e\}$$

$$D_i(\underline{Y}) = X_i.$$

## *Bit error Probability*

$$P_{b,i} \triangleq \mathbb{P}[D_i(\underline{Y}) \neq X_i].$$

$$P_b \triangleq \frac{1}{N} \sum_{i=1}^N P_{b,i}.$$



$$\mathbb{P}(Y_i = e) = p$$

**Observation 1**, *Bit error probability = Transinformation:*

$$P_{b,i} = H(X_i | \underline{Y}) \quad , \text{if } X_i \text{ is } \text{Ber}(1/2).$$

**Proof:**

if  $X_i$  can be recovered from  $\underline{Y} = \underline{y}$ ,  $H(X_i | \underline{Y} = \underline{y}) = 0$ .

Else,  $\mathbb{P}(X_i = x | \underline{Y} = \underline{y}) = 1/2$  and  $H(X_i | \underline{Y} = \underline{y}) = 1$ .

# MAP EXIT functions...to capture bit error probability

## **Definition:**

the vector EXIT function associated with bit  $i$ ,

$$h_i(\underline{p}) \triangleq H(X_i | \underline{Y}_{\sim i}(\underline{p}_{\sim i})).$$

the average EXIT function,

$$h(\underline{p}) \triangleq \frac{1}{N} \sum_{i=1}^N h_i(\underline{p})$$

we can choose  $p_i=p$  for all  $i$ .

## **Implication:**

It is the Transinformation associated with each bit ' $i$ ', which comes from all bits of received vector except bit ' $i$ ' (hence the name *EXtrinsic Information Transfer*)

## **Observation 2, Bit error probability vs EXIT function:**

By definition of conditional Entropy...

$$\begin{aligned} H(X_i | \underline{Y}) &= \mathbb{P}(Y_i = e) H(X_i | \underline{Y}_{\sim i}, Y_i = e) + \mathbb{P}(X_i = Y_i) H(X_i | \underline{Y}_{\sim i}, Y_i = X_i) \\ &= \mathbb{P}(Y_i = e) H(X_i | \underline{Y}_{\sim i}). \end{aligned}$$

Further using Observation 1,

$$P_{b,i}(p) = p h_i(p)$$

# MAP EXIT function ...area theorem

## **Proposition 1\*.**

*The MAP EXIT function satisfies ...*

$$h_i(p) = \frac{\partial H(\underline{X}/\underline{Y}(p))}{\partial p_i}$$

*Proof:*

$$\begin{aligned} H(\underline{X}/\underline{Y}(p)) &= H(X_i/\underline{Y}(p)) + H(X_{\sim i}/X_i, \underline{Y}(p)) \\ &= H(X_i/\underline{Y}(p)) + H(X_{\sim i}/X_i, Y_{\sim i}) \text{ ,by memorylessness} \\ &= p_i h_i(p) + H(X_{\sim i}/X_i, Y_{\sim i}) \end{aligned}$$

*noting that the second term is independent of  $p_i$ , and differentiating gives the above result.*

\* Note the propositions are numbered according to original paper

# MAP EXIT function ...area theorem

## **Proposition 4(c), Area Theorem:**

*The area under average EXIT function is equal to the rate*

$$\int_0^1 h(p) dp = \frac{K}{N}.$$

## **Implication:**

*This theorem will govern where the threshold will lie.*

*Proof:*

from proposition 1,  $h_i(p) = \frac{\partial H(\underline{X}|\underline{Y}(\underline{p}))}{\partial p_i}$

this implies, for a parameterized vector,  $\underline{p}(t) = (p_1(t), \dots, p_n(t))$ ,  $t \in (0, 1)$

$$H(\underline{X}|\underline{Y}(\underline{p}(1))) - H(\underline{X}|\underline{Y}(\underline{p}(0))) = \int_0^1 \left( \sum_{i=1}^N h_i(\underline{p}(t)) p'_i(t) \right) dt.$$

consider  $p_i(t) = t$

noting,  $H(\underline{X}|\underline{Y}(\underline{1})) = H(\underline{X}) = K$ ,  $H(\underline{X}|\underline{Y}(\underline{0})) = 0$ , and using the definition of  $h(p)$  completes the proof.

# MAP EXIT function ...as a measure of a set

Consider **bit erasure pattern**  $\mathbf{a}$  as a indicator vector s.t.  $\mathbf{a}_i = \mathbf{1}_{\{Y_i=e\}}$   
 A binary vector  $\mathbf{a}$  covers another vector  $\mathbf{c}$  if  $\mathbf{a}_i > \mathbf{c}_i$  for all  $i$ .

- A transmitted codeword can be recovered iff the erasure pattern does not cover any codeword.
- A bit  $i$  can be recovered uniquely iff the error pattern does not cover any codeword where bit  $i$  is non-zero.

	1	2	3	4	5	6	E does not cover
E	1	0	0	1	0	0	
C1	1	0	1	1	1	0	True
C2	1	1	0	1	0	1	True
C3	1	0	0	1	0	0	True
C4	1	0	0	0	0	0	X



# MAP EXIT function ...as a measure of a set

**Indirect Recovery:** Consider a code  $C$  and the indirect recovery of  $X_i$  from the subvector  $\underline{Y}_{\sim i}$  (i.e., the bit-MAP decoding of  $Y_i$  from  $\underline{Y}$  when  $Y_i = e$ ).

For  $i \in [N]$ , the **set of erasure patterns that prevent indirect recovery of  $X_i$**  under bit-MAP decoding is given by,

$$\Omega_i \triangleq \{A \subseteq [N] \setminus \{i\} : \exists B \subseteq [N] \setminus \{i\}, B \cup \{i\} \in \mathcal{C}, B \subseteq A\}.$$

**Proposition 3(a), 4(a),**  $h_i(p)$  is measure of  $\Omega_i$  :

$$h_i(p) = \mu_p(\Omega_i) = \sum_{A \in \Omega_i} p^{|A|} (1-p)^{N-1-|A|}.$$

Intuitively, it is the set of all erasure patterns, that cover some codeword with bit ' $i$ ' = 1, if ' $i$ ' is included. hence strengthens  $P_{b,i}(p) = p h_i(p)$

# MAP EXIT function ...as a measure of a set

*Proof:*

*by definition of conditional entropy,*

$$\begin{aligned} h_i(\underline{p}) &= H(X_i | \underline{Y}_{\sim i}(\underline{p}_{\sim i})) \\ &= \sum_{\underline{y}_{\sim i} \in \mathcal{Y}^{N-1}} \Pr(\underline{Y}_{\sim i} = \underline{y}_{\sim i}) H(X_i | \underline{Y}_{\sim i} = \underline{y}_{\sim i}). \end{aligned}$$

*WOLOG assume all 0's were sent. due to error pattern  $A$ , few bits got erased.*

*thus,*

$$\Pr(\underline{Y}_{\sim i} = \underline{y}_{\sim i}) = \prod_{\ell \in A} p_\ell \prod_{\ell \in A^c \setminus \{i\}} (1 - p_\ell).$$

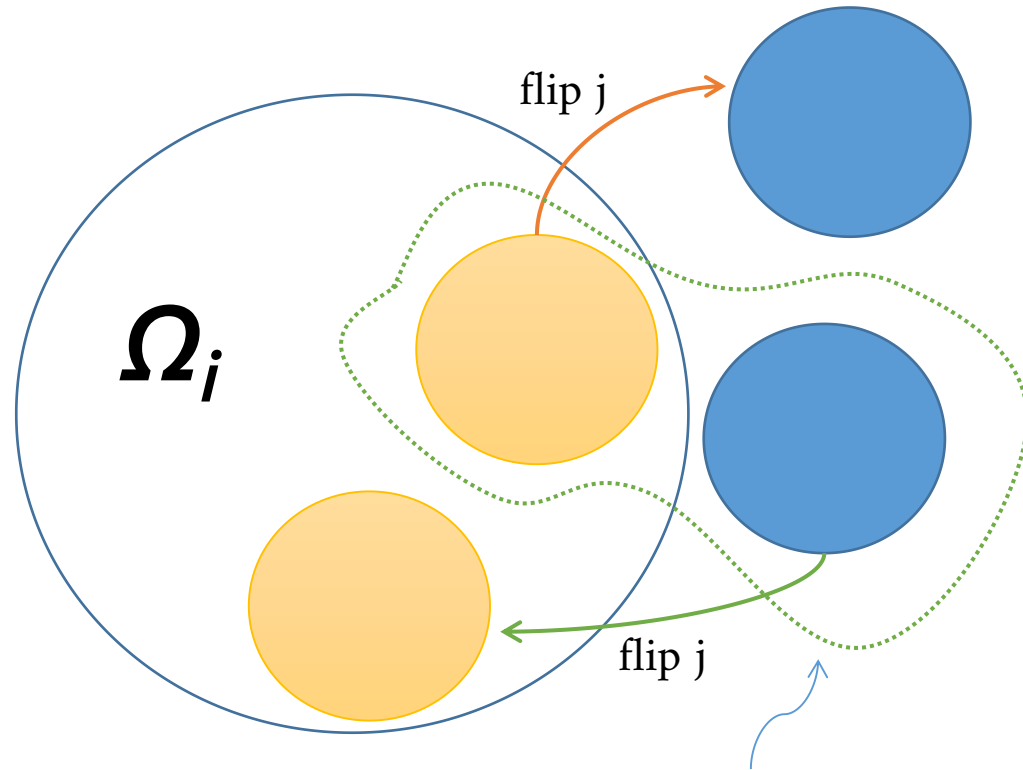
*Now, if  $A \in \Omega_i$ , then decoding fails hence,  $H(X_i | \underline{Y}_{\sim i} = \underline{y}_{\sim i}) = 1$ , else 0.*

*thus,*

$$h_i(\underline{p}) = \sum_{A \in \Omega_i} \prod_{\ell \in A} p_\ell \prod_{\ell \in A^c \setminus \{i\}} (1 - p_\ell).$$

*the result follows on considering,  $p_i = p$ .*

# MAP EXIT function ...Influences of $\Omega_i$



*The measure of this set is the  $j$ -th influence of  $\Omega_i$*

$$\partial_j \Omega_i \triangleq \{A \subseteq [N] \setminus \{i\} \mid A \setminus \{j\} \notin \Omega_i, A \cup \{j\} \in \Omega_i\}.$$

# MAP EXIT function ...Influences of $\Omega_i$

**Proposition 3(b),4(b)**,  $j$ th partial derivative of  $h_i(p)$  is  $j$ th influence of  $\Omega_i$  :

$$\frac{\partial^2 H(\underline{X}|\underline{Y}(\underline{p}))}{\partial p_j \partial p_i} = \frac{\partial h_i(\underline{p})}{\partial p_j} = \sum_{A \in \partial_j \Omega_i} \prod_{\ell \in A} p_\ell \prod_{\ell \in A^c \setminus \{i\}} (1 - p_\ell).$$

$$\left. \frac{\partial h_i(\underline{p})}{\partial p_j} \right|_{\underline{p}=(p,p,\dots,p)} = \mu_p(\partial_j \Omega_i) = \sum_{A \in \partial_j \Omega_i} p^{|A|} (1-p)^{N-1-|A|}.$$

*Proof: similar to proposition 3(a),4(a)*

We need to show next :

- all influences are equal, hence the set is symmetric ...
  - all  $h_i(p)$  are equal , hence area theorem can be applied...
- these will stem from the assumption that the code is 2-transitive

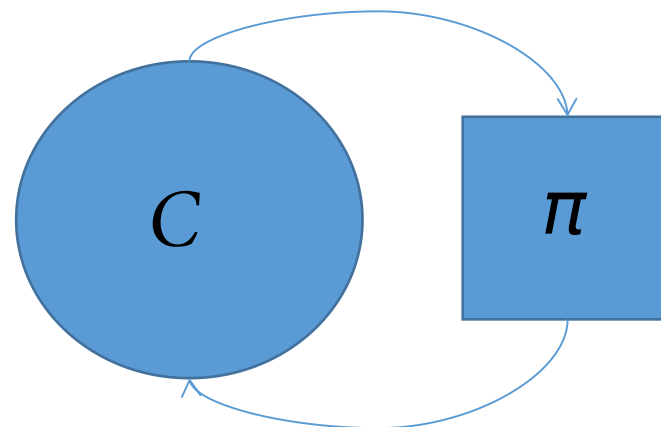
# Permutation group...briefing

- **Symmetric group  $S_N$  of  $[N]$ :**

*Is the set of all permutations of  $N$  possible.*

$[N=3]$  ,  $S_N = \{\{1,2,3\}, \{1,3,2\}, \{2,1,3\} \dots\}$

- **Permutation group** is a subset of symmetric group.



*Considering  $[N]$  denotes the set of indices of a code a permutation operation ( $\pi$ ) is a rearrangement of the bits of a code.*

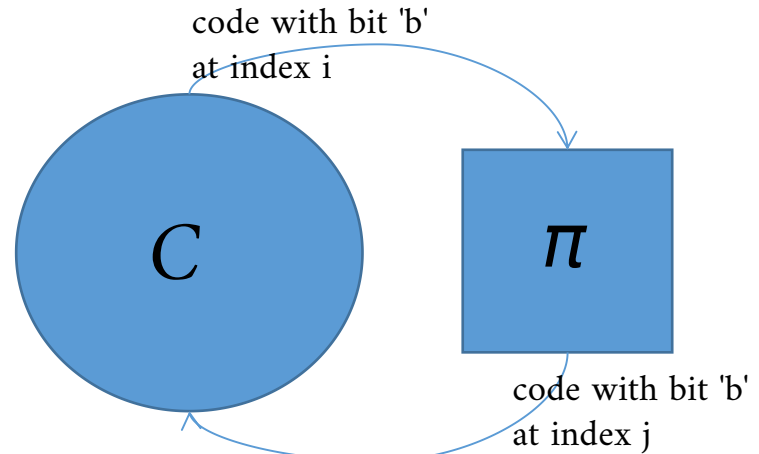
- A **Permutation automorphic group  $P_{Aut}(C)$**  of a code is the permutation group , such that under  $\pi \in P_{Aut}(C)$  the membership of the code is preserved, we denote  $P_{Aut}(C)$ , as

$$\mathcal{G} = \{\pi \in S_N : \pi(A) \in \mathcal{C} \text{ for all } A \in \mathcal{C}\}.$$

# Permutation group...transitivity

$$\mathcal{G} = \{\pi \in S_N : \pi(A) \in \mathcal{C} \text{ for all } A \in \mathcal{C}\}.$$

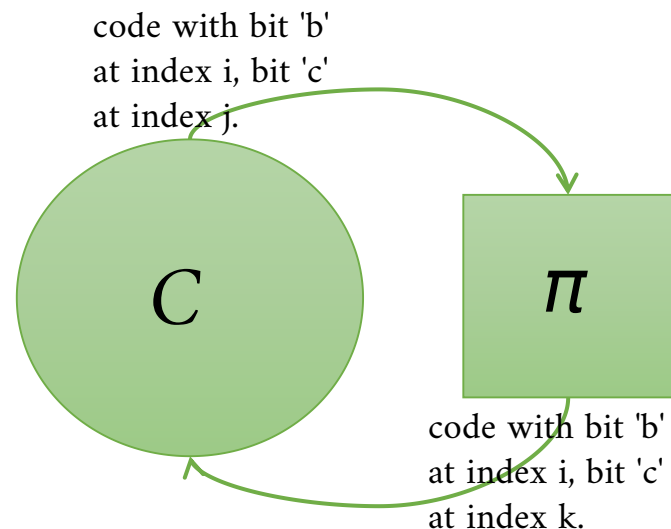
if for any  $i, j \in [N]$ , there exists,  
 $\pi \in \mathcal{G}$ , such that  $\pi(i) = j$ ,  
then  $\mathcal{G}$  is **transitive**.



$$\mathcal{G} = \{\pi \in S_N : \pi(A) \in \mathcal{C} \text{ for all } A \in \mathcal{C}\}.$$

if for any  $i, j, k \in [N]$ , there exists,  
 $\pi \in \mathcal{G}$ , such that,  
 $\pi(i) = i$  and  $\pi(j) = k$ .

then  $\mathcal{G}$  is **doubly-transitive**.



# MAP EXIT function ...for doubly transitive code

**Proposition 7**, All EXIT functions are equal , for a transitive code:

$$h(p) = h_i(p) \text{ for } 0 \leq p \leq 1.$$

*Proof.* claim: if  $\mathcal{G}$  is transitive, then so is  $\Omega_i$ .

As  $A \in \Omega_i$ , by definition  $\exists B$ , s.t.,  $B \cup \{i\} \in \mathcal{C}$ , but by transitivity of  $\mathcal{G}$ ,  $\pi(B \cup \{i\}) \in \mathcal{C}$ .

Observe,

$$\pi(B \cup \{i\}) = \pi(B) \cup \pi(\{i\}) = \pi(B) \cup j$$

.

Since  $\pi(B) \subseteq \pi(A)$ , it follows  $\pi(A) \in \Omega_j$ . This indicates a bijection between  $\Omega_i$  and  $\Omega_j$ , i.e.,  $|\Omega_i| = |\Omega_j|$ . Moreover since,  $|A| = |\pi(A)|$ , proposition follows from proposition 4 (a)  $\square$

i.e, using, 
$$h_i(p) = \mu_p(\Omega_i) = \sum_{A \in \Omega_i} p^{|A|} (1-p)^{N-1-|A|}.$$

# MAP EXIT function ...symmetry

**Proposition 8**, *All Influences are equal , for a transitive code:*

$$\left. \frac{\partial h_i(\underline{p})}{\partial p'_j} \right|_{\underline{p}=(p,p,\dots,p)} = \left. \frac{\partial h_i(\underline{p})}{\partial p'_k} \right|_{\underline{p}=(p,p,\dots,p)} .$$

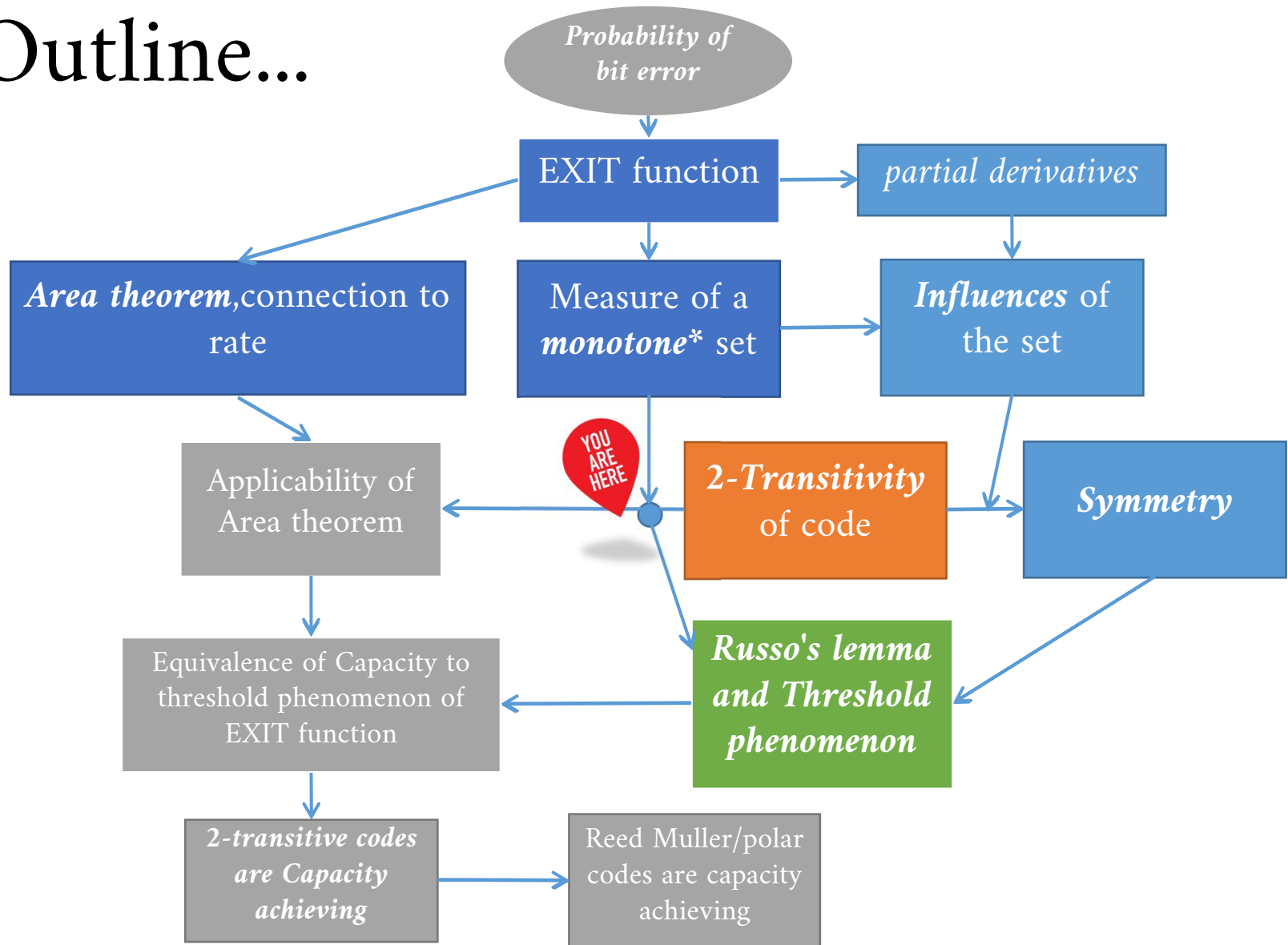
for distinct,  $i, j, k \in [N]$ ,

*Proof:*

*Similar to Proposition 7. Intuitively we expect that once we permute the locations , the bits which were pivotal must continue to remain so. Otherwise we could have decoded the concerned bit using simple permutations.*



# Outline...



# Capacity achieving codes...equivalence

Suppose  $\{C_n\}$  is a sequence of codes with rates  $\{r_n\}$  where  $r_n \rightarrow r$  for  $r \in (0, 1)$ . a)  $\{C_n\}$  is said to be **capacity achieving on the BEC** under bit-MAP decoding, if for any  $p \in [0, 1 - r)$ , the average bit-erasure probabilities satisfy,

$$\lim_{n \rightarrow \infty} P_b^{(n)}(p) = 0.$$

*The following are equivalent,*

- S1:  $\{C_n\}$  is capacity achieving on the BEC under bit-MAP decoding.
- S2: The sequence of average EXIT functions satisfies

$$\lim_{n \rightarrow \infty} h^{(n)}(p) = \begin{cases} 0 & \text{if } 0 \leq p < 1 - r \\ 1 & \text{if } 1 - r < p \leq 1. \end{cases}$$

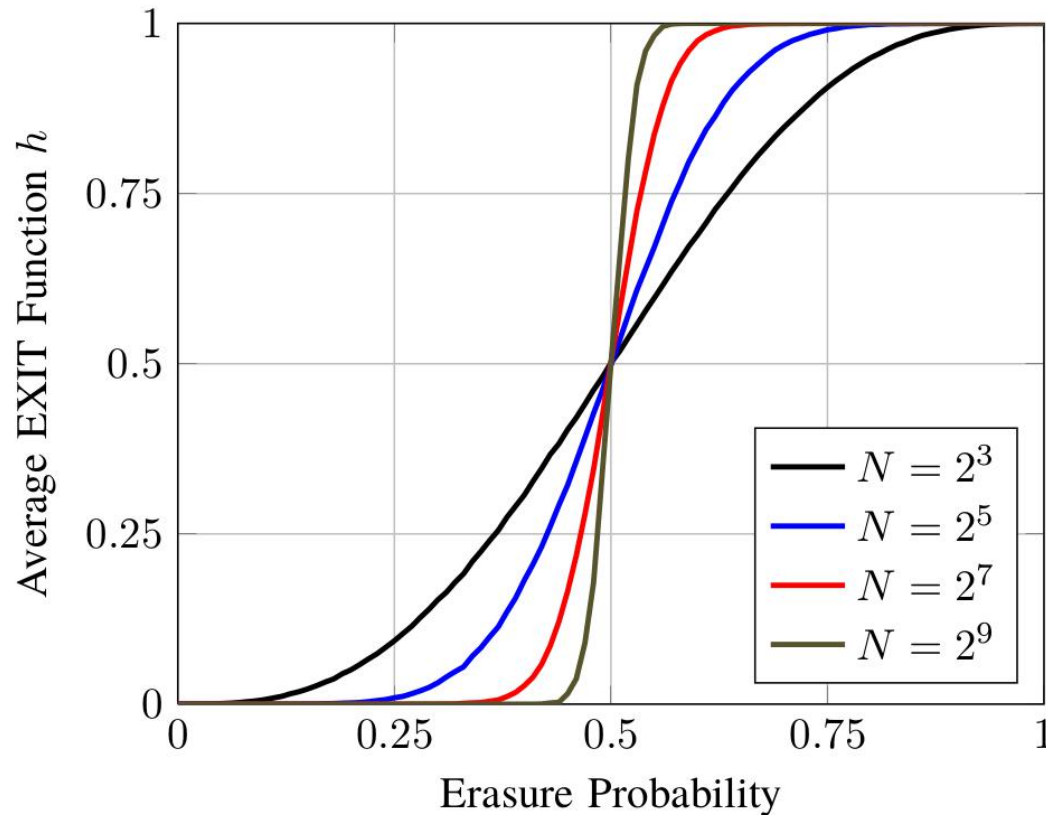
- S3: For any,

$$\lim_{n \rightarrow \infty} p_{1-\epsilon}^{(n)} - p_{\epsilon}^{(n)} = 0.$$

note,  $h^{(n)}(p_{\epsilon}^{(n)}) = \epsilon$ .

While S1  $\iff$  S2, follows from relationship between  $h_i(p)$  and probability of bit error, the rest follow from area theorem

# Capacity achieving codes...



*The average EXIT function of Rate-1/2 RM-code with Block length  $N$ . Shows threshold phenomenon.*

# Main result...threshold phenomenon

## **Influence:**

for a Monotone set  $\Omega$ , the  $j$ -th influence is defined as.

$$I_j^{(p)}(\Omega) \triangleq \mu_p(\partial_j \Omega)$$

the total, Influence is defined as,

$$I^{(p)} \triangleq \sum_{j=1}^N I_j^{(p)}.$$

**Theorem 19:** Let  $\Omega$  be a monotone set and suppose that, for all  $0 \leq p \leq 1$ , the influences of all bits are equal ,i.e,

$$I_1^{(p)}(\Omega) = \dots = I_M^{(p)}(\Omega).$$

Then, for any  $0 < \epsilon \leq 1/2$ ,

$$p_{1-\epsilon} - p_\epsilon \leq \frac{2 \log \frac{1-\epsilon}{\epsilon}}{C \log(N-1)},$$

where  $p_t = \inf\{p \in [0, 1] : \mu_p(\Omega) > t\}$  is well defined because  $\mu_p(\Omega)$  is strictly increasing in  $p$  with  $\mu_0(\Omega) = 0$  and  $\mu_1(\Omega) = 1$ .

*Proof:* follows from Russo's lemma

# Main result...*notations*

## ***Redefinitions***

*We can redefine  $\Omega_i$  as a set of indicator vectors of  $A$ .*

$$[\phi_i(A)]_l = \begin{cases} \mathbf{1}_A(l) & \text{if } l < i \\ \mathbf{1}_A(l+1) & \text{if } l \geq i. \end{cases}$$

$$\begin{aligned} \Omega'_i &\triangleq \{\phi_i(A) \in \{0,1\}^{N-1} : A \in \Omega_i\} \\ \partial_j \Omega'_i &\triangleq \{\phi_i(A) \in \{0,1\}^{N-1} : A \in \partial_j \Omega_i\}. \\ &= \{\underline{x} \in \{0,1\}^{N-1} | \mathbb{1}_{\Omega_i}(\underline{x}) \neq \mathbb{1}_{\Omega_i}(\underline{x}^{(j)})\} \end{aligned}$$

*Accordingly ,we can refine the definition of the measure...*

$$\mu_p(\Omega) = \sum_{\underline{x} \in \Omega} p^{|\underline{x}|} (1-p)^{M-|\underline{x}|}, \text{ for } \Omega \subseteq \{0,1\}^M,$$

# Main result...

Let  $\{C_n\}$  be a sequence of codes where the blocklengths satisfy  $N_n \rightarrow \infty$ , the rates satisfy  $r_n \rightarrow r$ , and the permutation group  $G_{(n)}$  (of  $C_n$ ) is **doubly transitive** for each  $n$ . If  $r \in (0, 1)$ , **then  $\{C_n\}$  is capacity achieving on the BEC under bit-MAP decoding.**

*Proof:*

Note,  $h(p) = h_i(p) = \mu_p(\Omega'_i)$  , where  $h^{(n)}(p)$  is the EXIT function of  $C_n$ .

## **Properties of $\Omega'_i$**

- $\Omega'_i$  is the set of error patterns that prevent the detection of  $X_i$ .  
Consider  $A \in \Omega'_i$  , any  $B > A$  will surely cause error. Hence  $B \in \Omega'_i$  and thus,  **$\Omega'_i$  is monotone.**

- Recall,  $I_j^p(\Omega'_i) = \mu_p(\partial_j \Omega'_i) = \frac{\partial h_i(\underline{p})}{\partial p'_j} \Big|_{\underline{p}=(p,p,\dots,p)}$

hence,  $I_j^p(\Omega'_i) = I_k^p(\Omega'_i)$  for all  $j, k \in [N-1]$ .  **$\Omega'_i$  is symmetric.**

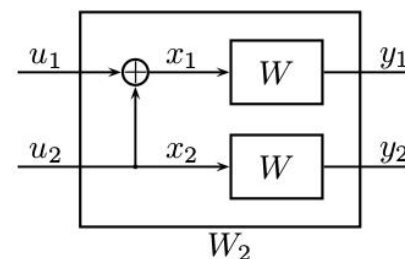
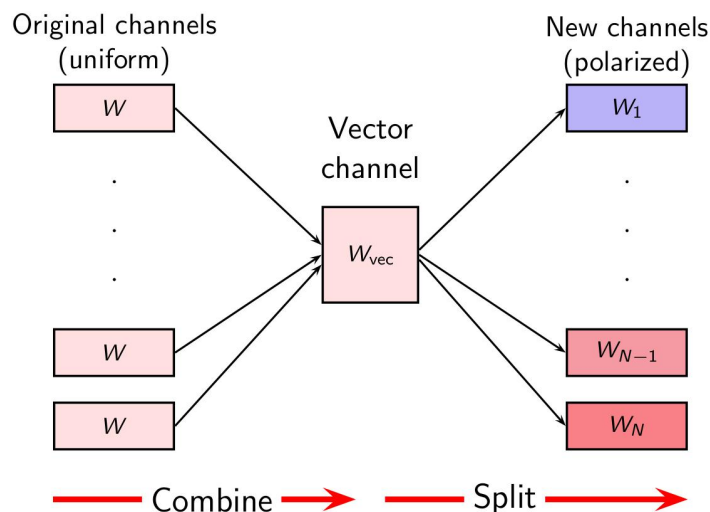
This implies,  $p_{1-\epsilon} - p_\epsilon \leq \frac{2 \log \frac{1-\epsilon}{\epsilon}}{C \log(N-1)}$ ,

Hence  $\lim_{n \rightarrow \infty} (p_{1-\epsilon} - p_\epsilon) = 0$  . , and  **$C_n$  achieves Capacity.**

# Polar codes are Capacity achieving...

## Introduction to polar codes

Channel polarization is an operation by which one manufactures out of  $N$  independent copies of a given B-DMC  $W$ , a second set of  $N$  channels  $\{W^{(i)}_N : 1 \leq i \leq N\}$  that show a polarization effect in the sense that, as  $N$  becomes large, the symmetric capacity terms  $\{I(W^{(i)}_N)\}$  **tend towards 0 or 1**, for all but a vanishing fraction of indices  $i$ . This operation consists of a channel combining and a channel splitting phase.



$$x_1^N = u_1^N G_N$$

# Polar codes are Capacity achieving...

*polar codes are doubly - transitive ,hence achieve capacity*

*Proof:*

*Assume we are given four distinct locations in the code  $a, b, c$  and  $d \in [N]$  ,we need to give a permutation  $\Pi : [N] \rightarrow [N]$  such that,*

*$\Pi(a) = c$  and  $\Pi(b) = d$  , and it is automorphic.*

*Consider any permutation which satisfies the above constraint,proving it is automorphic suffices.*

$$\text{Note , } \Pi_{N \times N} * [u_1 \dots u_n] = [u_{\Pi(1)} \dots u_{\Pi(n)}]$$

$$\text{and, } \Pi_{N \times N} * [x_1 \dots x_n] = [x_{\Pi(1)} \dots x_{\Pi(n)}]$$

*Hence,upon premultiplying the encoding equation with  $\Pi$  ,we get*

$$\Pi * [x_1 \dots x_n] = \Pi * [u_1 \dots u_n] * G_n$$

$$\Rightarrow [x_{\Pi(1)} \dots x_{\Pi(n)}] = [u_{\Pi(1)} \dots u_{\Pi(n)}] * G_n$$

*As polar codes are linear , and  $[u_{\Pi(1)} \dots u_{\Pi(n)}]$  is a valid message, so  $[x_{\Pi(1)} \dots x_{\Pi(n)}]$  is a valid codeword too.hence proved*





# Thanks...