

Analysis of Channel Polarization

A Theoretical Motivation towards Polar Codes

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Background knowledge

- Probability Theory and Concentration Inequalities
- Information Theory
- Martingale Random Processes

The Original Paper/Reference

Erdal Arıkan, *Channel Polarization: A Method for Constructing Capacity-Achieving Codes for Symmetric Binary-Input Memoryless Channels*, IEEE Transactions On Information Theory, Vol. 55, No. 7, July 2009

Definitions/Notation

$$W : \mathcal{X} \rightarrow \mathcal{Y} \quad W(y|x), x \in \mathcal{X}, y \in \mathcal{Y}.$$

W is a transition probability map or **channel**.

$$I(W) \triangleq \sum_{y \in \mathcal{Y}} \sum_{x \in \mathcal{X}} \frac{1}{2} W(y|x) \log \frac{W(y|x)}{\frac{1}{2} W(y|0) + \frac{1}{2} W(y|1)}$$

$I(W)$ is the symmetric capacity of the channel. Here, x is uniformly distributed over $\{0, 1\}$. This parameter is strongly tied to the rate of transmission. $0 \leq I(W) \leq 1$.

$$Z(W) \triangleq \sum_{y \in \mathcal{Y}} \sqrt{W(y|0)W(y|1)}.$$

$Z(W)$ is the Bhattacharya parameter of the channel. It measures the “reliability” of the channel. $0 \leq Z(W) \leq 1$.

$$W^N : \mathcal{X}^N \rightarrow \mathcal{Y}^N$$

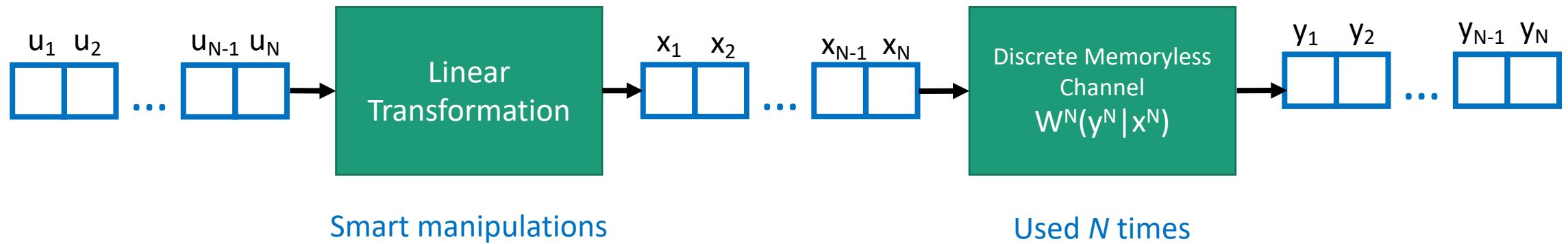
$$W^N(y^N|x^N) = \prod_{i=1}^N W(y_i|x_i)$$

W^N is the equivalent channel when W is used N times independently.

$$\begin{aligned} y^N &:= y_1^N := (y_1, y_2, \dots, y_N), \\ x^N &:= x_1^N := (x_1, x_2, \dots, x_N) \\ x_a^b &:= (x_a, \dots, x_b) \end{aligned}$$

Channel Polarization

- Use $W(y|x)$ independently N times and **artificially manufacture** a new set of channels $W_N^{(i)}$ which are **polarized**, i.e., $I(W_N^{(i)})$ goes to either 0 or 1, $\forall i$ asymptotically.



- Channel polarization can be visualized by breaking down the entire operation to two phases
 1. Channel Combining
 2. Channel Splitting
- For decoding, we look at
 3. Successive Cancellation

Channel Polarization

1. Channel Combining:

Consider the following definition of a channel $W_N : \mathcal{X}^N \rightarrow \mathcal{Y}^N, W_N(y^N|x^N)$ by using independent copies of $W(y|x)$. $N = 2^k$.

$$W_1(y_1|u_1) = W(y_1|u_1)$$

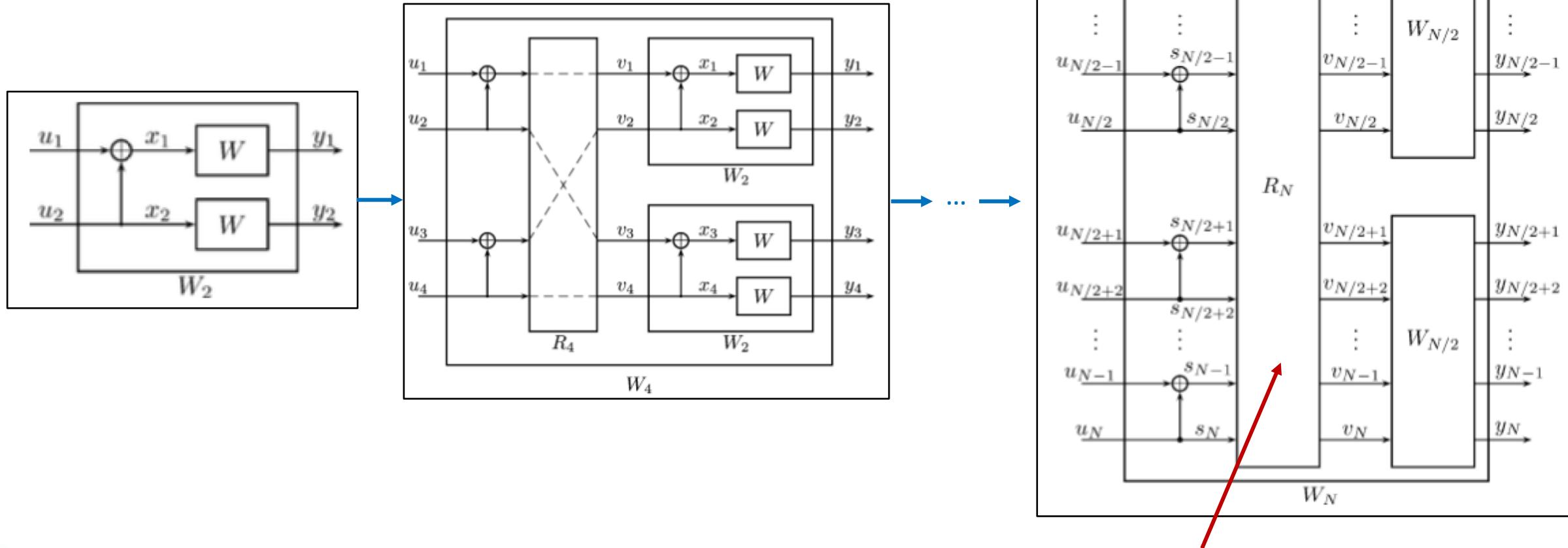
$$W_2(y_1, y_2|u_1, u_2) = W_1(y_1|u_1 \oplus u_2)W_1(y_2|u_2)$$

$$W_4(y_1, y_2, y_3, y_4|u_1, u_2, u_3, u_4) = W_2(y_2, y_1|u_1 \oplus u_2, u_3 \oplus u_4)W_2(y_3, y_4|u_2, u_4)$$

...

Channel Polarization

1. Channel Combining:



Reverse shuffle operation: Group all odd terms in the first half and all even terms in the second half sequentially.

Recursive Construction of W_N

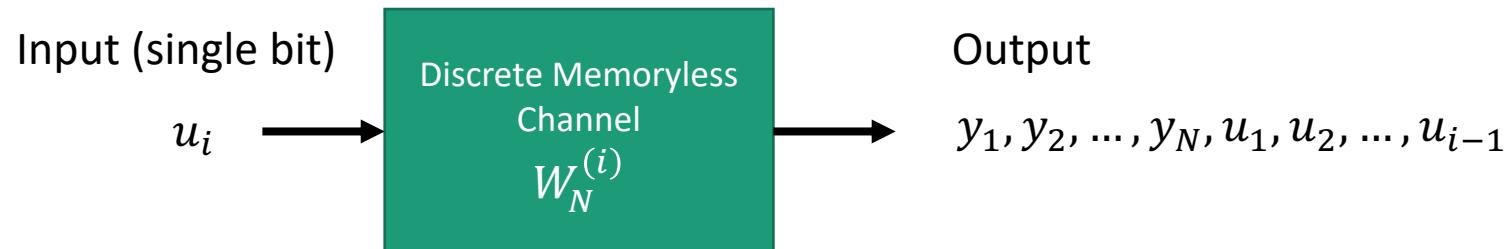
Channel Polarization

2. Channel Splitting:

We now split W_N channel into a set of individual single bit input channels $\bar{W}_N^{(i)} : \mathcal{X} \rightarrow \mathcal{Y}^N \times \mathcal{X}^{i-1}$, $1 \leq i \leq N$ defined as

$$W_N^{(i)}(y_1^N, u_1^{i-1} | u_i) \triangleq \sum_{u_{i+1}^N \in \mathcal{X}^{N-i}} \frac{1}{2^{N-1}} W_N(y_1^N | u_1^N)$$

Summed over all
“unknown” bits



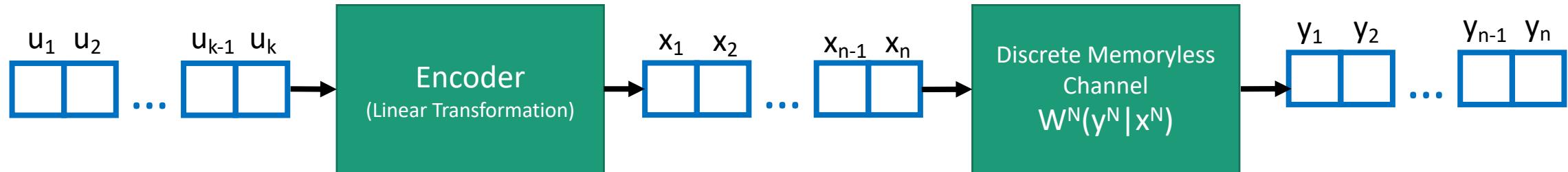
What we will try to do:

Given $y_1, y_2, \dots, y_N, u_1, u_2, \dots, u_{i-1}$, try estimating u_i .

We will prove that this channel is polarized, i.e., its capacity is very close to either 1 or 0.

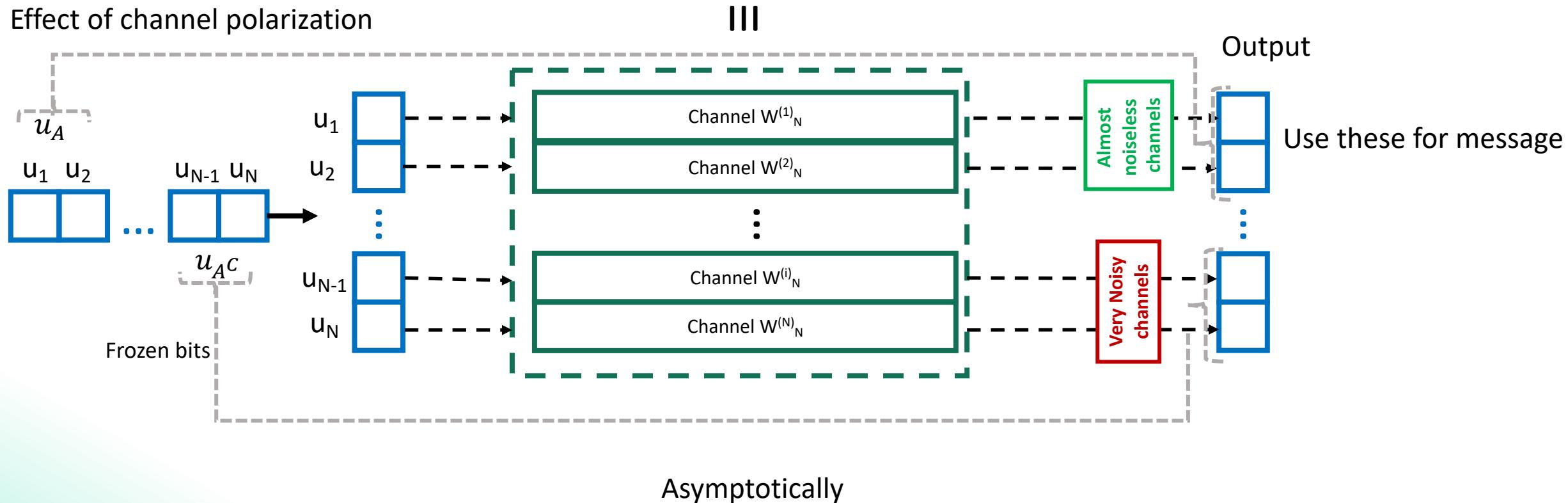
Polar codes: Intuition

Transmission with coding



Channel is used N times (N large)

Effect of channel polarization



Decoding

3. Successive Cancellation Decoder

$$\hat{u}_i \triangleq \begin{cases} u_i, & \text{if } i \in \mathcal{A}^c \\ h_i(y_1^N, \hat{u}_1^{i-1}), & \text{if } i \in \mathcal{A} \end{cases}$$

$$h_i(y_1^N, \hat{u}_1^{i-1}) \triangleq \begin{cases} 0, & \text{if } \frac{W_N^{(i)}(y_1^N, \hat{u}_1^{i-1}|0)}{W_N^{(i)}(y_1^N, \hat{u}_1^{i-1}|1)} \geq 1 \\ 1, & \text{otherwise} \end{cases}$$

Proofs

Single step transformation of (W, W) to (W', W'')

Definition:

Consider two independent copies of channel W . The following is a single step transformation

$$(W, W) \mapsto (W', W'')$$

Iff $\forall u_1, u_2 \in \mathcal{X}, y_1, y_2 \in \mathcal{Y}$

$$W'(y_1, y_2 | u_1) = \frac{1}{2} \sum_{u'_2} W(y_1 | u_1 + u'_2) W(y_2 | u'_2)$$

$$W''(y_1, y_2, u_1 | u_2) = \frac{1}{2} W(y_1 | u_1 + u_2) W(y_2 | u_2)$$

From this definition, we see $(W, W) \rightarrow (W_2^{(1)}, W_2^{(2)})$

$$W_N^{(i)}(y_1^N, u_1^{i-1} | u_i) \triangleq \sum_{u_{i+1}^N \in \mathcal{X}^{N-i}} \frac{1}{2^{N-1}} W_N(y_1^N | u_1^N)$$

Can be generalized to $(W_N^{(i)}, W_N^{(i)}) \rightarrow (W_{2N}^{(2i-1)}, W_{2N}^{(2i)})$ **(L1)**

L1. Recursive Construction of $W_N^{(i)}$

$$\begin{aligned} & W_{2N}^{(2i-1)}(y_1^{2N}, u_1^{2i-2} | u_{2i-1}) \\ &= \sum_{u_{2i}} \frac{1}{2} W_N^{(i)}(y_1^N, u_{1,o}^{2i-2} \oplus u_{1,e}^{2i-2} | u_{2i-1} \oplus u_{2i}) \\ & \quad \cdot W_N^{(i)}(y_{N+1}^{2N}, u_{1,e}^{2i-2} | u_{2i}) \end{aligned}$$

$$\begin{aligned} & W_{2N}^{(2i)}(y_1^{2N}, u_1^{2i-1} | u_{2i}) \\ &= \frac{1}{2} W_N^{(i)}(y_1^N, u_{1,o}^{2i-2} \oplus u_{1,e}^{2i-2} | u_{2i-1} \oplus u_{2i}) \\ & \quad \cdot W_N^{(i)}(y_{N+1}^{2N}, u_{1,e}^{2i-2} | u_{2i}) \end{aligned}$$

i.e., $(W_N^{(i)}, W_N^{(i)}) \rightarrow (W_{2N}^{(2i-1)}, W_{2N}^{(2i)})$

L2. Channel Parameters after Single step transforms

If $(W, W) \rightarrow (W', W'')$, then

$$\begin{aligned} I(W') + I(W'') &= 2I(W) \\ I(W') &\leq I(W'') \end{aligned}$$

Proof:

Let U_1, U_2 be the inputs to the channels. Define $X_1 = U_1 + U_2$ and $X_2 = U_2$.

We send (X_1, X_2) , linear transformation of input vector (U_1, U_2) , through the channel W^2

$$I(W') = I(Y_1, Y_2; U_1)$$

$$I(W'') = I(Y_1, Y_2, U_1; U_2) = I(U_1; U_2) + I(Y_1, Y_2; U_2|U_1) = I(Y_1, Y_2; U_2|U_1)$$

$$I(W') + I(W'') = I(Y_1, Y_2; U_1) + I(Y_1, Y_2; U_2|U_1) = I(Y_1, Y_2; U_1, U_2) = I(Y_1, Y_2; X_1, X_2)$$

$$\therefore I(W') + I(W'') = 2I(W)$$

Chain Rule of Mutual Information

$$\begin{aligned} \text{Also, } I(W'') &= I(Y_1, Y_2, U_1; U_2) = I(Y_2; U_2) + I(Y_1, U_1; U_2|Y_2) \geq I(W) \\ \Rightarrow I(W') &\leq I(W) \end{aligned}$$

$$\begin{aligned} W'(y_1, y_2|u_1) &= \frac{1}{2} \sum_{u'_2} W(y_1|u_1 + u'_2)W(y_2|u'_2) \\ W''(y_1, y_2, u_1|u_2) &= \frac{1}{2} W(y_1|u_1 + u_2)W(y_2|u_2) \end{aligned}$$

L2. Channel Parameters after Single step transforms

If $(W, W) \rightarrow (W', W'')$, then

$$Z(W'') = Z(W)^2$$

$$Z(W') \leq 2Z(W) - Z(W)^2$$

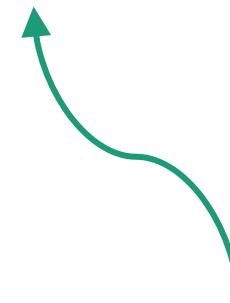
$$Z(W') \geq Z(W) \geq Z(W'').$$

Proof: First Equality

$$\begin{aligned} Z(W'') &= \sum_{y_1, y_2, u_1} \sqrt{W''(y_1, y_2, u_1|0)W''(y_1, y_2, u_1|1)} = \frac{1}{2} \sum_{y_1, y_2, u_1} \sqrt{W(y_1|u_1)W(y_2|0)W(y_1|u_1+1)W(y_2|1)} \\ &= \frac{1}{2} \sum_{y_1, u_1} \sqrt{W(y_1|u_1)W(y_1|u_1+1)} \sum_{y_2} \sqrt{W(y_2|0)W(y_2|1)} \\ &= \frac{1}{2} \times (2Z(W)) \times (Z(W)) \\ &= Z(W)^2 \\ \therefore Z(W'') &\leq Z(W) \end{aligned}$$

Second Equality can be shown to be true with simple algebraic identities.

Third inequality can be shown by exploiting the convex property of $Z(W)$ and Minkowski's Inequality.



$$W'(y_1, y_2|u_1) = \frac{1}{2} \sum_{u'_2} W(y_1|u_1 + u'_2)W(y_2|u'_2)$$

$$W''(y_1, y_2, u_1|u_2) = \frac{1}{2} W(y_1|u_1 + u_2)W(y_2|u_2)$$

Summary (Till Now)

$$\begin{aligned} W_{2N}^{(2i-1)}(y_1^{2N}, u_1^{2i-2} | u_{2i-1}) \\ = \sum_{u_{2i}} \frac{1}{2} W_N^{(i)}(y_1^N, u_{1,o}^{2i-2} \oplus u_{1,e}^{2i-2} | u_{2i-1} \oplus u_{2i}) \\ \cdot W_N^{(i)}(y_{N+1}^{2N}, u_{1,e}^{2i-2} | u_{2i}) \end{aligned}$$

$$\begin{aligned} W_{2N}^{(2i)}(y_1^{2N}, u_1^{2i-1} | u_{2i}) \\ = \frac{1}{2} W_N^{(i)}(y_1^N, u_{1,o}^{2i-2} \oplus u_{1,e}^{2i-2} | u_{2i-1} \oplus u_{2i}) \\ \cdot W_N^{(i)}(y_{N+1}^{2N}, u_{1,e}^{2i-2} | u_{2i}) \end{aligned}$$

Using **L1** and **L2**, we can summarize

A

$$(W_N^{(i)}, W_N^{(i)}) \rightarrow (W_{2N}^{(2i-1)}, W_{2N}^{(2i)})$$

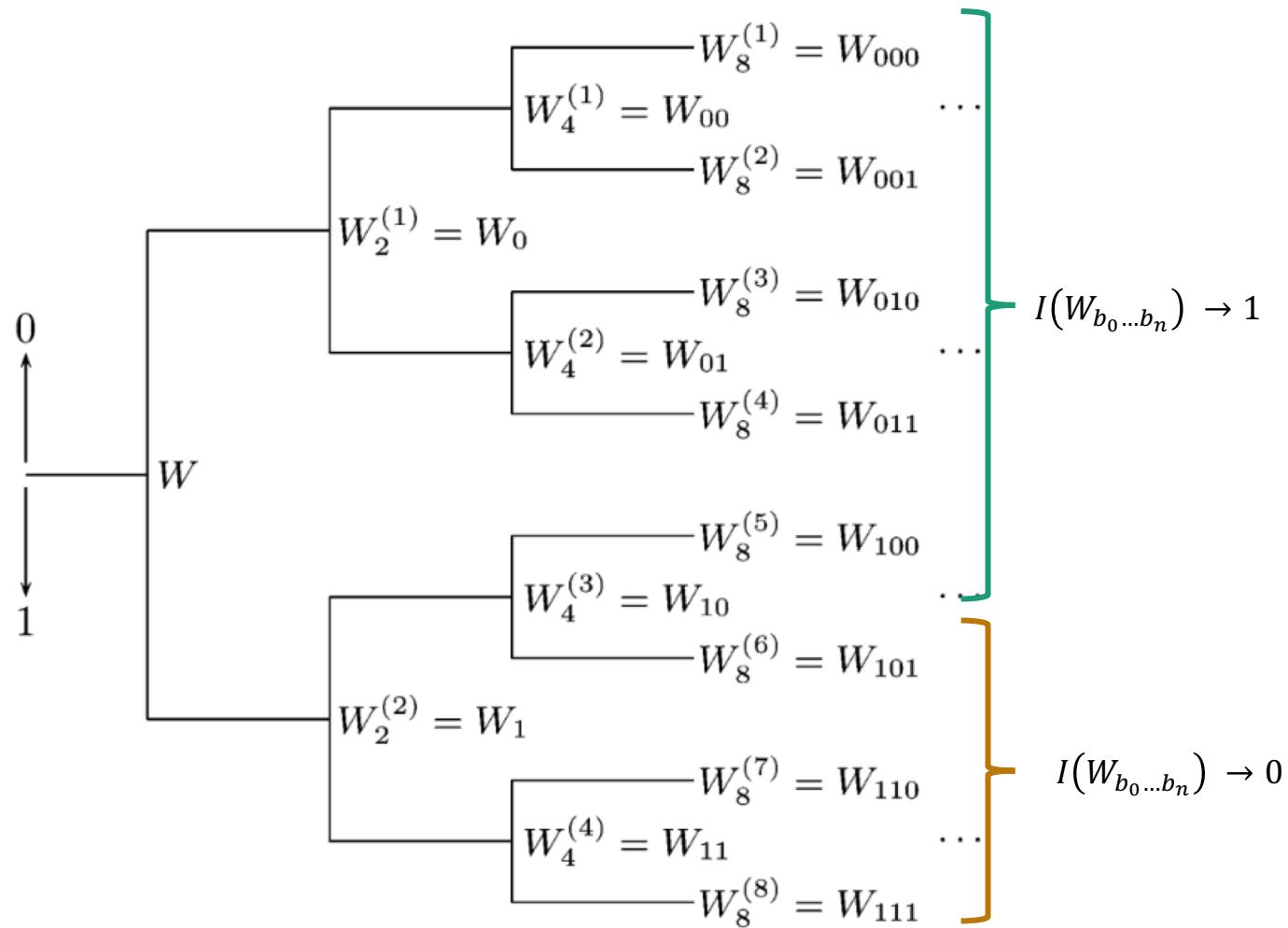
B

$$\begin{aligned} I(W_{2N}^{(2i-1)}) + I(W_{2N}^{(2i)}) &= 2I(W_N^{(i)}) \\ I(W_{2N}^{(2i-1)}) &\leq I(W_N^{(i)}) \leq I(W_{2N}^{(2i-1)}) \end{aligned}$$

C

$$\begin{aligned} Z(W_{2N}^{(2i)}) &= Z(W_N^{(i)})^2 \\ Z(W_{2N}^{(2i-1)}) + Z(W_{2N}^{(2i)}) &\leq 2Z(W_N^{(i)}) \end{aligned}$$

Binary/Power of 2 recursion



$$(W_N^{(i)}, W_N^{(i)}) \rightarrow (W_{2N}^{(2i-1)}, W_{2N}^{(2i)})$$

Since $N = 2^n$, $W_N^{(i)} = W_{b_0, b_1, \dots, b_n}$ where

$$i = 1 + \sum_{r=0}^{n-1} b_r 2^r$$

T1. Channel Polarization

For any B-DMC W , the channels $\{W_N^{(i)}\}$ polarize, i.e., for any $\delta > 0$, as $N = 2^n$ goes to ∞ , a subset of indices $A \subset \{1, \dots, N\}$, $i \in A, j \in A^c$, $I(W_N^{(i)}) \in (1 - \delta, 1]$ and $I(W_N^{(j)}) \in [0, \delta)$ with $\frac{|A|}{N} = I(W)$.

Proof:

Define $\{b_n\}_{n \geq 0}$ to be an i.i.d. Bernoulli random process, such that $b_n = 0$ or 1 with equal probability $\frac{1}{2}$.

Then, $W_{b_0 b_1 \dots b_n}$ is a random process defined on the tree in the previous figure, with $W_0 = W$, the true B-DMC.

Moreover, $I_n \triangleq I(W_{b_0 b_1 \dots b_n})$ and $Z_n \triangleq Z(W_{b_0 b_1 \dots b_n})$ are defined random processes.

$$E[I_n | b^{n-1}] = E[I(W_{b_0 b_1 \dots b_n}) | b^{n-1}] = \frac{1}{2}I(W_{b_0 b_1 \dots b_{n-1} 0}) + \frac{1}{2}I(W_{b_0 b_1 \dots b_{n-1} 1}) = I(W_{b_0 b_1 \dots b_{n-1}})$$

$$\text{because } I(W_{2N}^{(2i-1)}) + I(W_{2N}^{(2i)}) = 2I(W_N^{(i)})$$

Hence I_n is a **bounded martingale process**, as $0 \leq I_n \leq 1$.

All moments of I_n exist!

From Martingale convergence, we have that I_∞ is a well defined random variable and $E|I_\infty - I_n| < \infty$.

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Proof:

$$E[Z_n | b^{n-1}] = E[Z(W_{b_0 b_1 \dots b_n}) | b^{n-1}] = \frac{1}{2} Z(W_{b_0 b_1 \dots b_{n-1} 0}) + \frac{1}{2} Z(W_{b_0 b_1 \dots b_{n-1} 1}) \leq Z(W_{b_0 b_1 \dots b_{n-1}})$$

because $Z(W_{2N}^{(2i-1)}) + Z(W_{2N}^{(2i)}) \leq 2Z(W_N^{(i)})$.

Hence Z_n is a **bounded supermartingale process**, as $0 \leq Z_n \leq 1$.

All moments of Z_n exist!

From Martingale convergence, we have that Z_∞ is a well defined random variable and $E|Z_\infty| < \infty$.
Since $Z_n = \sum_{i=1}^n (Z_i - Z_{i-1})$, and as $E|Z_n|$ converges, we have $E|Z_{n+1} - Z_n| \rightarrow 0$.

But Z_{n+1} is $Z(W_{b_0 b_1 \dots b_n 0}) = Z_n^2$ with probability $\frac{1}{2}$ because $Z(W_{2N}^{(2i)}) = Z(W_N^{(i)})^2$.

Hence, $E|Z_{n+1} - Z_n| \geq \frac{1}{2} E[Z_n^2 - Z_n] = \frac{1}{2} E[Z_n(1 - Z_n)]$.

T1. Channel Polarization

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Proof:

$$\text{Hence, } E|Z_{n+1} - Z_n| \geq \frac{1}{2}E[|Z_n^2 - Z_n|] = \frac{1}{2}E[|Z_n(1 - Z_n)|].$$

As $E|Z_{n+1} - Z_n| \rightarrow 0$, we also have $E[|Z_n(1 - Z_n)|] \rightarrow 0$, which implies Z_n converges to either 0 or 1 almost surely!

As $Z_\infty = 0$ or 1, we have $I_\infty = 1 - Z_\infty$. (Can see intuitively that $Z(W) = 0$ gives $I(W) = 1$ and vice-versa)

The above result is true whenever $Z(W) = 0$ or 1.

$$Z(W) \triangleq \sum_{y \in \mathcal{Y}} \sqrt{W(y|0)W(y|1)}.$$

But since I_n is a martingale, $E[I_\infty] = I_0$ which immediately gives us

$$P(I_\infty = 1) = I_0 = I(W) \text{ and } P(I_\infty = 0) = 1 - I_0$$



T2. Channel Polarization

For any B-DMC W , and any fixed rate $R < I(W)$, there exists a sequence of sets $A_N \subseteq \{1, \dots, N\}$, $N = 2^n$ such that $|A_N| \geq NR$ and $Z(W_N^{(i)}) \leq O(N^{-\frac{5}{4}}) \forall i \in A_N$.

Proof:

From the same setting as earlier, we have

$$\begin{cases} Z_{n+1} \leq Z_n^2, & b_n = 1 \quad [\text{as } Z(W_{2N}^{(2i)}) = Z(W_N^{(i)})^2] \\ Z_{n+1} \leq 2Z_n - Z_n^2 \leq 2Z_n, & b_n = 0 \quad [\text{as } Z(W_{2N}^{(2i-1)}) \leq 2Z(W_N^{(i)}) - Z(W_N^{(i)})^2] \end{cases}$$

For parameters $2 \geq \zeta \geq 0, m \geq 0$

$$T_m(\zeta) \triangleq \{\omega \in \Omega; Z_i \leq \zeta, \forall i \geq m\}$$

Then for $\omega \in T_m(\zeta)$ and $i \geq m$, we have

$$\frac{Z_{i+1}}{Z_i} \leq \begin{cases} 2, & b_n = 0 \\ \zeta, & b_n = 1 \end{cases}$$

This implies $Z_n \leq \zeta 2^{n-m} \prod_{i=m+1}^n \left(\frac{\zeta}{2}\right)^{b_i} = \zeta 2^{n-m} \left(\frac{\zeta}{2}\right)^{\sum_{i=m+1}^n b_i}$ for $n > m$

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Proof:

For $n > m \geq 0$ and $0 < \eta < 1/2$, define,

$$U_{m,n}(\eta) \triangleq \left\{ \omega \in \Omega : \sum_{i=m+1}^n b_i \geq \left(\frac{1}{2} - \eta\right)(n-m) \right\}$$

Then

$$Z_n(\omega) \leq \zeta \left[2^{\frac{1}{2}+\eta} \zeta^{\frac{1}{2}-\eta} \right]^{n-m}, \quad \omega \in T_m(\zeta) \cap U_{m,n}(\eta)$$

Substitute $\eta_0 = 1/20$ and $\zeta_0 = 2^{-4}$ to get

$$Z_n(\omega) \leq 2^{-4 - \frac{5(n-m)}{4}}, \quad \omega \in T_m(\zeta_0) \cap U_{m,n}(\eta_0)$$

We need to show that for a given m, n , $T_m(\zeta_0) \cap U_{m,n}(\eta_0)$ occurs with high probability.

T2. Channel Polarization

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Proof:

First consider $T_m(\zeta_0)$.

As seen from **T1**, $P(Z_\infty = 0) = I_0$, which implies that $S_n := \{Z_n \leq \zeta_0\}$, $\{Z_\infty = 0\} \subseteq \cup_{n \geq m} S_n$ for large enough m , and hence $P(\cup_{n \geq m} S_n) \geq I_0 - \frac{\delta}{2}$.

$T_{m_0}(\zeta_0) = \cup_{n \geq m_0} S_n$, and from continuity of probability,

$$P(T_{m_0}(\zeta_0)) = P(\cup_{n \geq m_0} S_n) \geq I_0 - \frac{\delta}{2}, \quad m_0 = m_0(\zeta_0, \delta)$$

Now consider $U_{m,n}(\eta)$.

$$\begin{aligned} P(U_{m,n}^c(\eta_0)) &= P\left(\sum_{i=m+1}^n b_i < \left(\frac{1}{2} - \eta_0\right)(n-m)\right) = P\left(-t \sum_{i=m+1}^n b_i > -t\left(\frac{1}{2} - \eta_0\right)(n-m)\right) \\ &\leq 2^{t\left(\frac{1}{2} - \eta_0\right)(n-m)} E[2^{-tb_i}]^{n-m} = \left[\left(\frac{2^{t\left(\frac{1}{2} - \eta_0\right)} + 2^{-t\left(\frac{1}{2} + \eta_0\right)}}{2}\right)\right]^{n-m} \end{aligned}$$

Chernoff Bound

$$P(U_{m,n}^c(\eta_0)) \leq 2^{-(n-m)\left(1-H\left(\frac{1}{2} - \eta_0\right)\right)}$$

T2. Channel Polarization

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Proof:

$$P(T_{m_0}^c(\zeta_0)) \leq 1 - I_0 + \frac{\delta}{2}, \quad m_0 = m_0(\zeta_0, \delta)$$

$$P(U_{m,n}^c(\eta_0)) \leq 2^{-(n-m)\left(1-H\left(\frac{1}{2}-\eta_0\right)\right)}$$

We can choose a finite $n_0(m_0, \eta_0, \delta)$ such that the RHS of the above inequality becomes at most $\frac{\delta}{2}$. Hence, from union bound, $\forall n > n_0$

$$P(T_{m_0}^c(\zeta_0) \cup U_{m_0,n}^c(\eta_0)) \leq 1 - I_0 + \delta$$

$$P(T_{m_0}(\zeta_0) \cap U_{m_0,n}(\eta_0)) \geq I_0 - \delta$$

Therefore

$$Z_n(\omega) \leq 2^{-4 - \frac{5(n-m_0)}{4}} = c2^{-\frac{5n}{4}}, \quad \omega \in T_{m_0}(\zeta_0) \cap U_{m_0,n}(\eta_0), \forall n > n_0$$

T2. Channel Polarization

For any B-DMC W , and any fixed rate $R < I(W)$, there exists a sequence of sets $A_N \subseteq \{1, \dots, N\}$, $N = 2^n$ such that $|A_N| \geq NR$ and $Z(W_N^{(i)}) \leq O(N^{-\frac{5}{4}}) \forall i \in A_N$.

Proof:

Define $V_n = \left\{\omega: Z_n \leq c2^{-\frac{5n}{4}}\right\}$, then immediately $T_{m_0}(\zeta_0) \cap U_{m_0,n}(\eta_0) \subseteq V_n \forall n > n_0$
 $P(V_n) \geq I - \delta = R \forall n > 0$

Notice that

$$P(V_n) = \sum_{b_0, b_1, \dots, b_n} \frac{1}{2^n} \mathbf{1}_{\{Z_n \leq c2^{-\frac{5n}{4}}\}} = \frac{|A_N|}{N}$$

Therefore,

$$|A_N| \geq NR \text{ and } Z(W_N^{(i)}) \leq O(N^{-\frac{5}{4}}) \forall i \in A_N$$



T3. Probability of Decoding error

$$\hat{u}_i \triangleq \begin{cases} u_i, & \text{if } i \in \mathcal{A}^c \\ h_i(y_1^N, \hat{u}_1^{i-1}), & \text{if } i \in \mathcal{A} \end{cases}$$

$$h_i(y_1^N, \hat{u}_1^{i-1}) \triangleq \begin{cases} 0, & \text{if } \frac{W_N^{(i)}(y_1^N, \hat{u}_1^{i-1}|0)}{W_N^{(i)}(y_1^N, \hat{u}_1^{i-1}|1)} \geq 1 \\ 1, & \text{otherwise} \end{cases}$$

$$P(\mathcal{E}) \leq \sum_{i=1}^n Z\left(W_N^{(i)}\right) \leq O\left(N^{-\frac{1}{4}}\right), \text{ where, } \mathcal{E} \triangleq \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : \hat{U}_A(u^N, y^N) \neq u_A\} \text{ and } R < I(W).$$

Proof:

Define $B_i = \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : \hat{U}_1^{i-1} = u_1^{i-1} \text{ and } \hat{U}_i \neq u_i\}$. Then, $\mathcal{E} = \bigcup_{i \in A} B_i$.

$$\begin{aligned} B_i &= \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : \hat{U}_1^{i-1} = u_1^{i-1} \text{ and } h_i(y_1^N, \hat{U}_1^{i-1}) \neq u_i\} \\ &= \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : \hat{U}_1^{i-1} = u_1^{i-1} \text{ and } h_i(y_1^N, u_1^{i-1}) \neq u_i\} \\ &\subseteq \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : h_i(y_1^N, u_1^{i-1}) \neq u_i\} \subseteq \mathcal{E}_i \end{aligned}$$

$$\text{where } \mathcal{E}_i = \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : W_N^{(i)}(y_1^N, u_1^{i-1}|u_i) \leq W_N^{(i)}(y_1^N, u_1^{i-1}|u_i + 1)\}$$

Therefore,

$$P(\mathcal{E}) \leq P(\bigcup_{i \in A} \mathcal{E}_i) \leq \sum_{i \in A} P(\mathcal{E}_i)$$

We now need an upper bound on $P(\mathcal{E}_i)$.

T3. Probability of Decoding error

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$$h_i(y_1^N, \hat{u}_1^{i-1}) \triangleq \begin{cases} 0, & \text{if } \frac{W_N^{(i)}(y_1^N, \hat{u}_1^{i-1}|0)}{W_N^{(i)}(y_1^N, \hat{u}_1^{i-1}|1)} \geq 1 \\ 1, & \text{otherwise} \end{cases}$$

$$P(\mathcal{E}) \leq \sum_{i=1}^n Z\left(W_N^{(i)}\right) \leq O\left(N^{-\frac{1}{4}}\right), \text{ where, } \mathcal{E} \triangleq \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : \hat{U}_A(u^N, y^N) \neq u_A\} \text{ and } R < I(W).$$

Proof:

$$\begin{aligned} P(\mathcal{E}_i) &= \sum_{u^N, y^N} P_{U^N, Y^N}(u^N, y^N) \mathbf{1}_{\{(u^N, y^N) \in \mathcal{E}_i\}} = \sum_{u^N, y^N} \frac{1}{2^N} W_N(y^N | u^N) \mathbf{1}_{\{(u^N, y^N) \in \mathcal{E}_i\}} \\ &\leq \sum_{u^N, y^N} \frac{1}{2^N} W_N(y^N | u^N) \sqrt{\frac{W_N^{(i)}(y_1^N, u_1^{i-1} | u_i + 1)}{W_N^{(i)}(y_1^N, u_1^{i-1} | u_i)}} \quad \text{Since we know an error has occurred} \\ &= \sum_{u_1^{i-1}, y^N} \sum_{u_i} \frac{1}{2} \left(\sum_{u_{i+1}^N} \frac{1}{2^{N-1}} W_N(y^N | u^N) \right) \sqrt{\frac{W_N^{(i)}(y_1^N, u_1^{i-1} | u_i + 1)}{W_N^{(i)}(y_1^N, u_1^{i-1} | u_i)}} \\ &= \sum_{u_i} \frac{1}{2} \sum_{u_1^{i-1}, y^N} \sqrt{W_N^{(i)}(y_1^N, u_1^{i-1} | u_i) W_N^{(i)}(y_1^N, u_1^{i-1} | u_i + 1)} = Z\left(W_N^{(i)}\right) \end{aligned}$$

T3. Probability of Decoding error

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$P(\mathcal{E}) \leq \sum_{i=1}^n Z(W_N^{(i)}) \leq O(N^{-\frac{1}{4}})$, where, $\mathcal{E} \triangleq \{(u^N, y^N) \in \mathcal{X}^N \times \mathcal{Y}^N : \hat{U}_A(u^N, y^N) \neq u_A\}$ and $R < I(W)$.

Proof:

$$P(\mathcal{E}_i) \leq Z(W_N^{(i)})$$

Hence,

$$\begin{aligned} P(\mathcal{E}) &\leq \sum_{i \in A} P(\mathcal{E}_i) \leq \sum_{i \in A} Z(W_N^{(i)}) \\ &\leq |A| \max\left(Z(W_N^{(i)})\right) \\ &\leq N \max\left(Z(W_N^{(i)})\right) \\ &\leq O(N^{-\frac{1}{4}}) \end{aligned}$$



Conclusions

Conclusions

- Polar codes can be asymptotically rate achieving codes for B-DMCs.
- For a block length of $N = 2^n$ and transmission rate $R < I(W)$, we see $Z\left(W_N^{(i)}\right) \sim N^{-\frac{5}{4}}$, for at least NR channels. This gives us a rate at which channels polarize.

We can get better bounds and tighter bounds than this.

- For a block length of $N = 2^n$ and transmission rate $R < I(W)$, average probability of error goes to zero asymptotically, as, $P(\text{error}) \sim N^{-\frac{1}{4}}$.

This is a loose bound, and we can strengthen the upper bound to an exponential function of N .

The upper bound also does not explicitly depend on R , and one can try to obtain a sharper bound which tells us how the error probability degrades with increasing rate.