

# Polarization and polar codes

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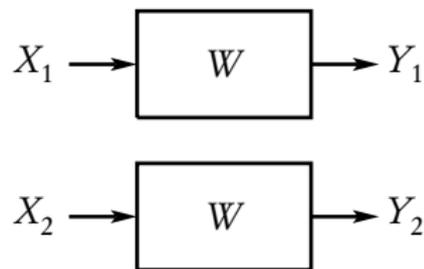
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- The technique is **information lossless**, and of **low complexity**\*
- I am **not** the inventor of this technique.

# Building block

Given two copies of a binary input channel  $W: \mathbb{F}_2 \rightarrow \mathcal{Y}$



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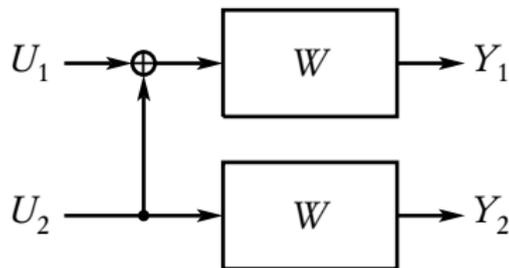
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$$X_1 = U_1 \oplus U_2$$

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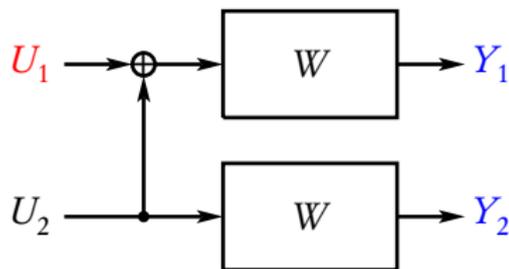
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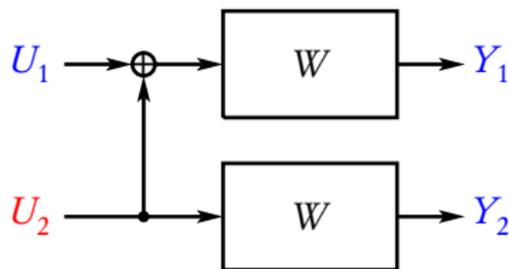
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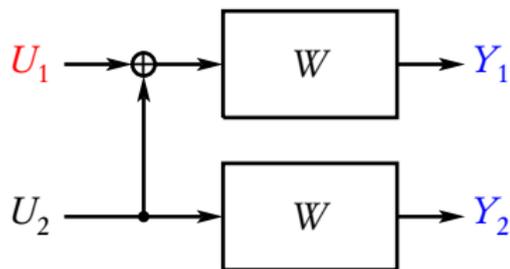
Note that

$$W^-(y_1, y_2 | u_1) = \sum_{u_2 \in \mathbb{F}_2} \frac{1}{2} W(y_1 | u_1 \oplus u_2) W(y_2 | u_2)$$
$$W^+(y_1, y_2, u_1 | u_2) = \frac{1}{2} W(y_1 | u_1 \oplus u_2) W(y_2 | u_2)$$

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Properties of  $W \mapsto (W^-, W^+)$ :

$$I(W^-) = I(U_1; Y_1 Y_2)$$

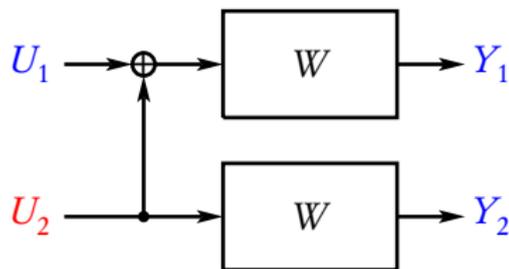


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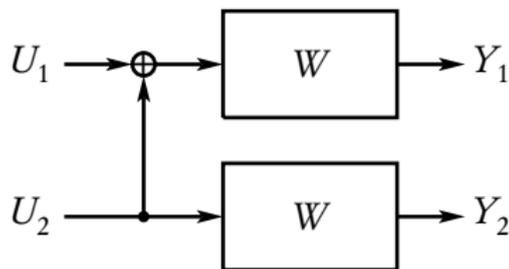
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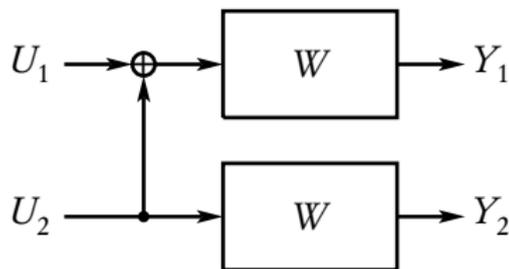
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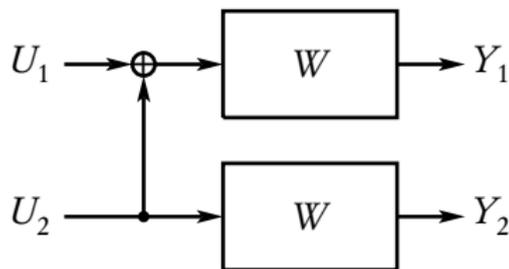
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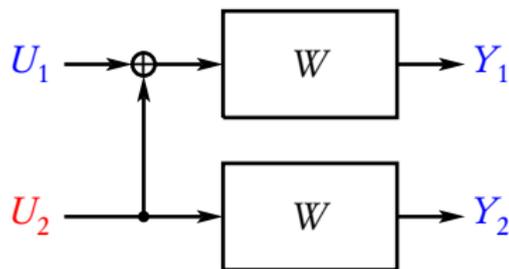
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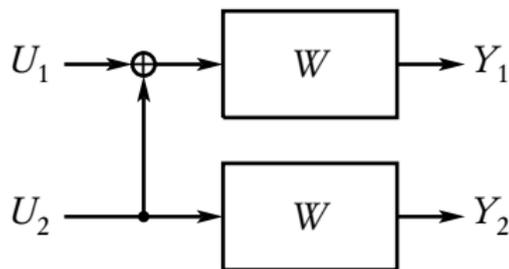
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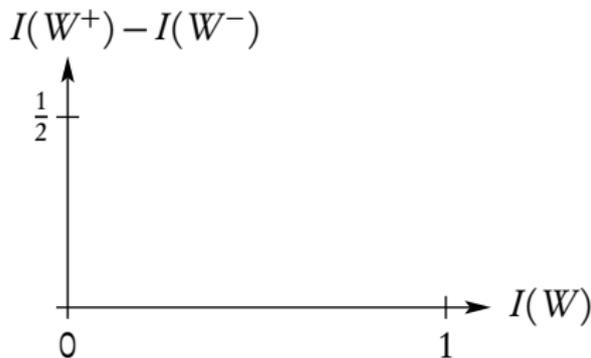
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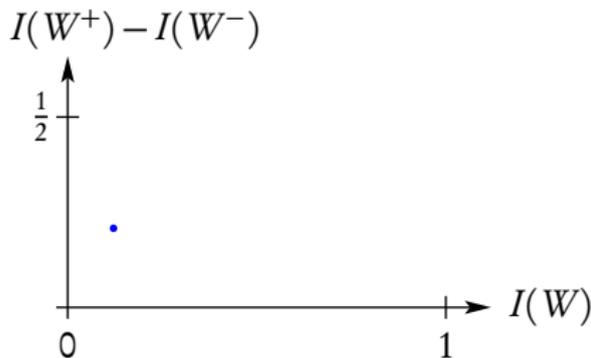
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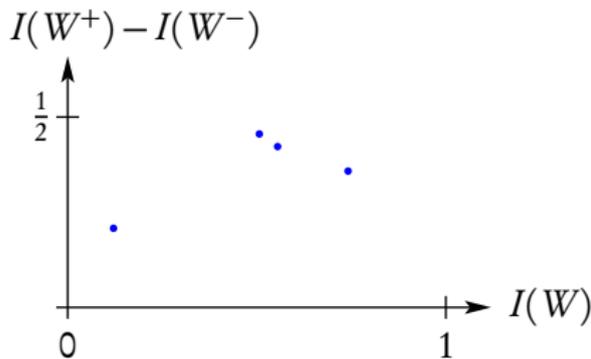
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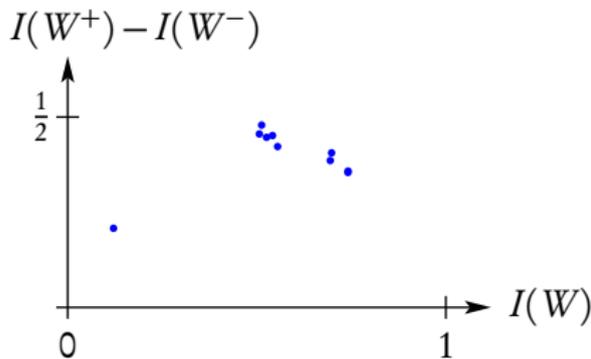
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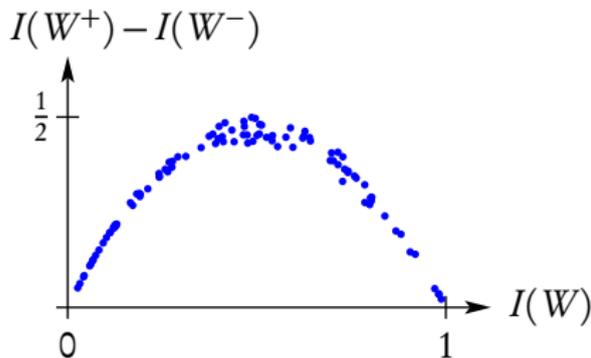
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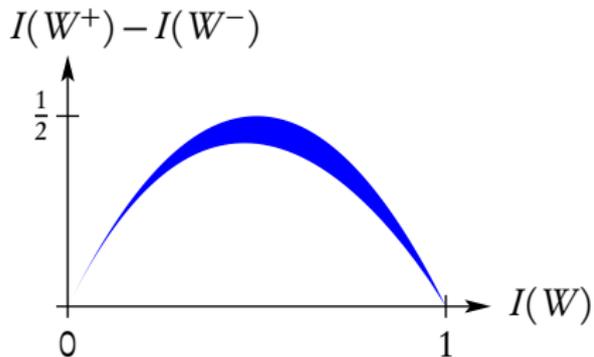
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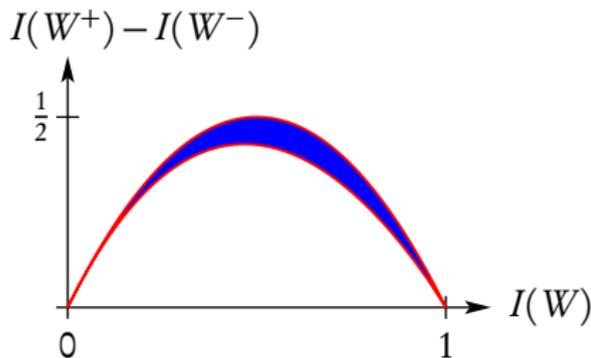
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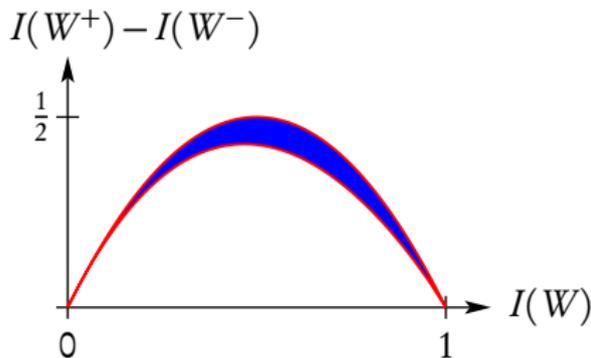
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- $I(W^+) \geq I(W) \geq I(W^-)$ .
- For every  $\epsilon > 0$  there is a  $\delta > 0$  such that

$I(W^+) - I(W^-) < \delta$  implies

$$I(W) \notin (\epsilon, 1 - \epsilon).$$



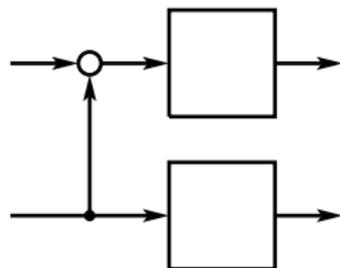
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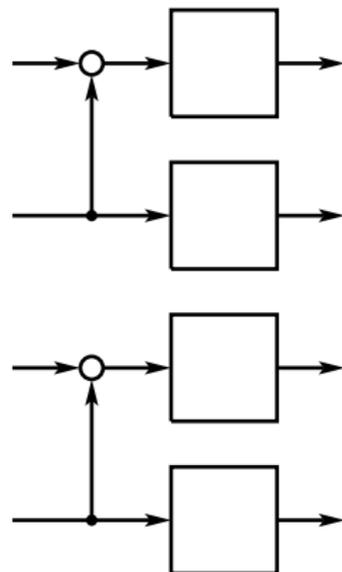
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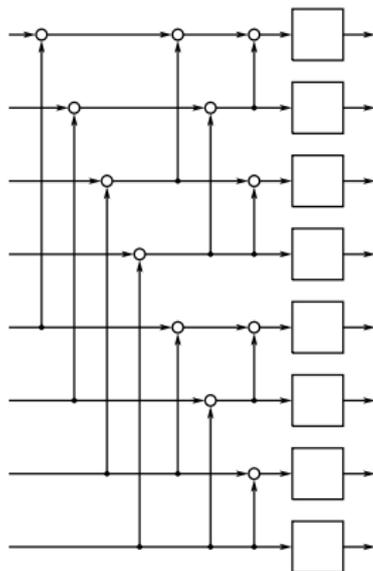




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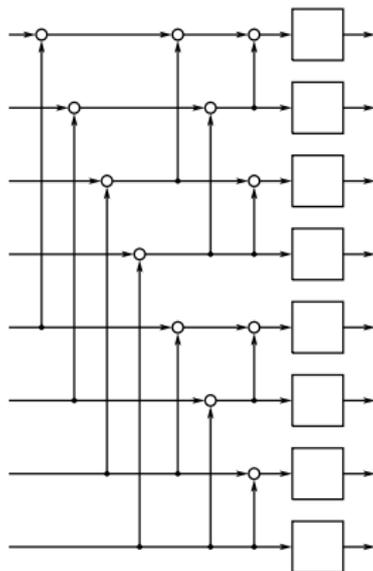
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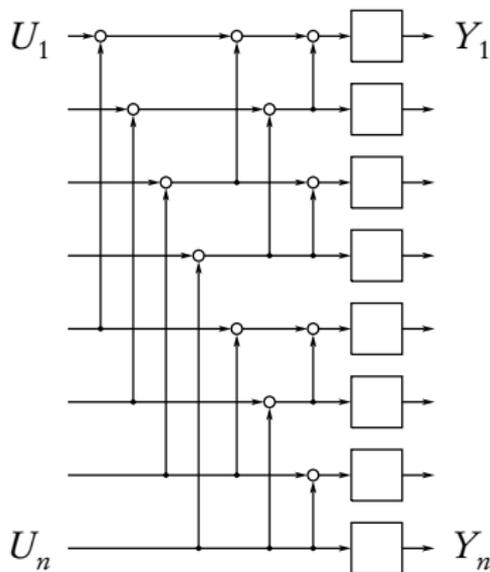
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- $\ell$  levels into this process, we have transformed  $n = 2^\ell$  uses of channel  $W$  to one use each of  $2^\ell$  channels

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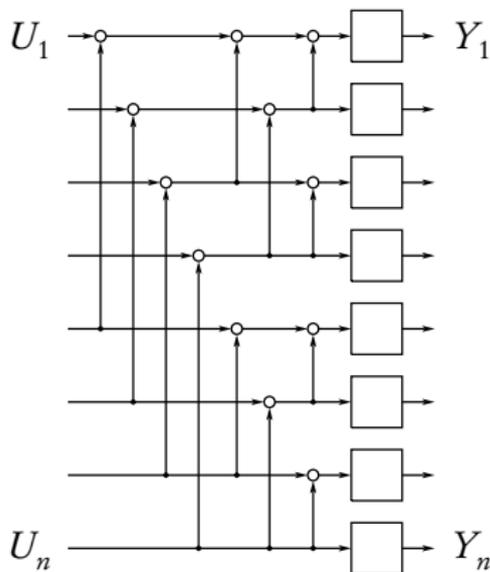
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- The quantities  $\{I(W^{b_1, \dots, b_\ell})\}$  are exactly the  $n$  quantities

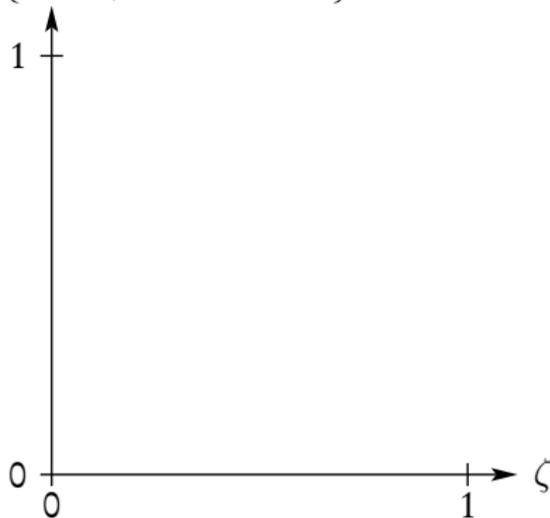
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# Example

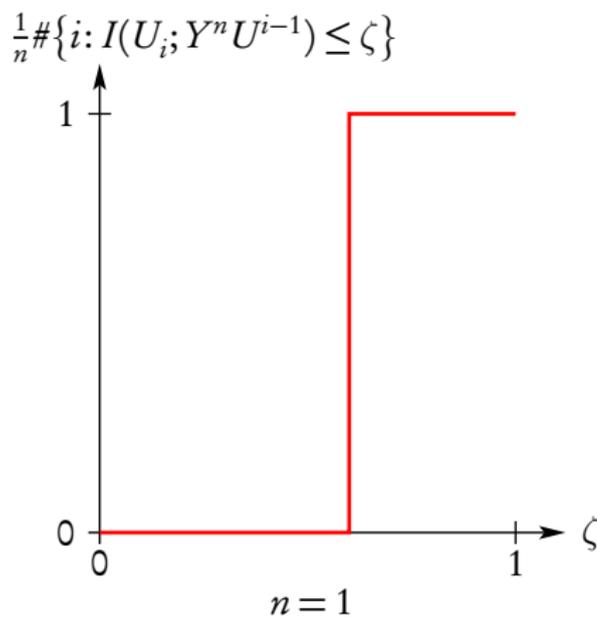
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$$\frac{1}{n} \#\{i: I(U_i; Y^n U^{i-1}) \leq \zeta\}$$



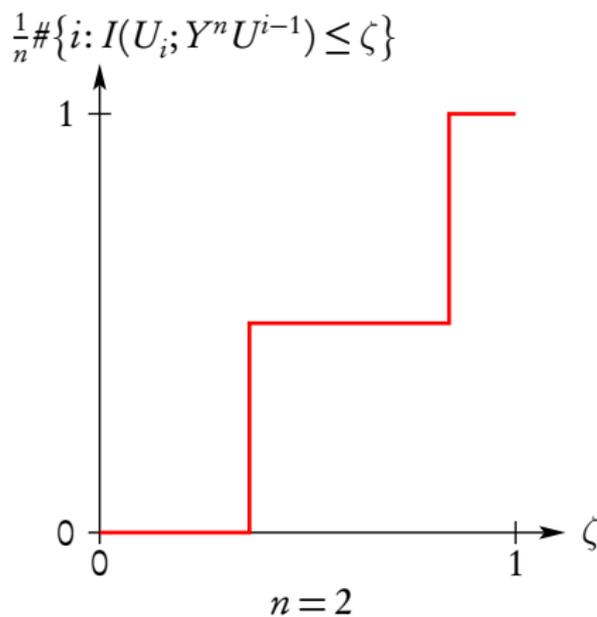
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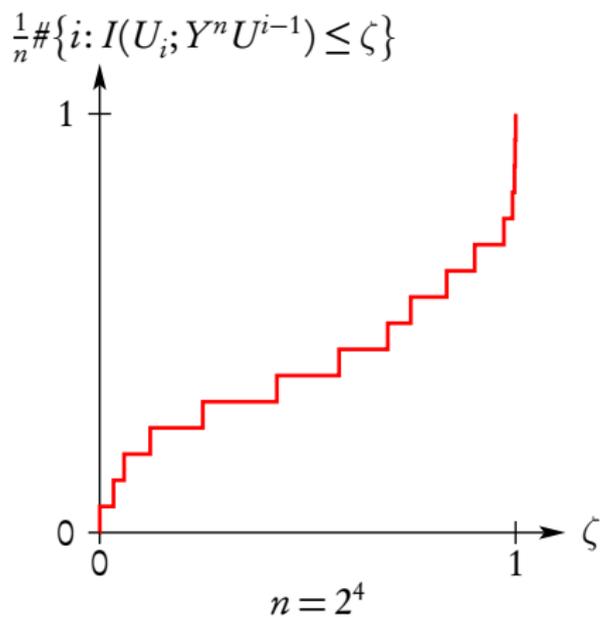
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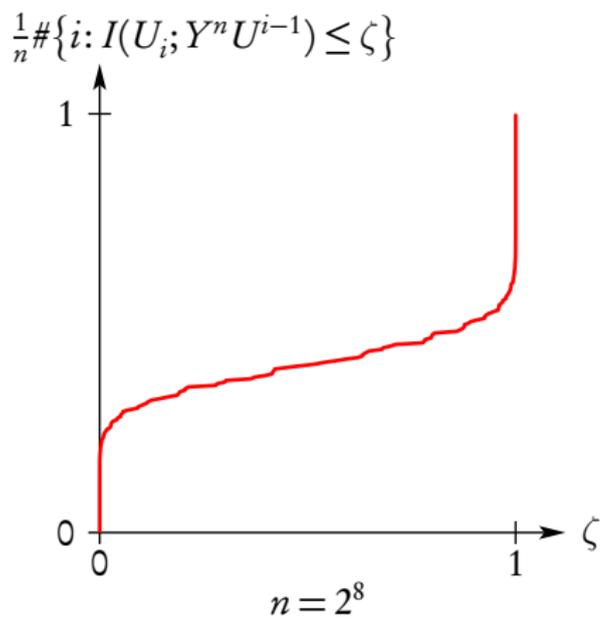
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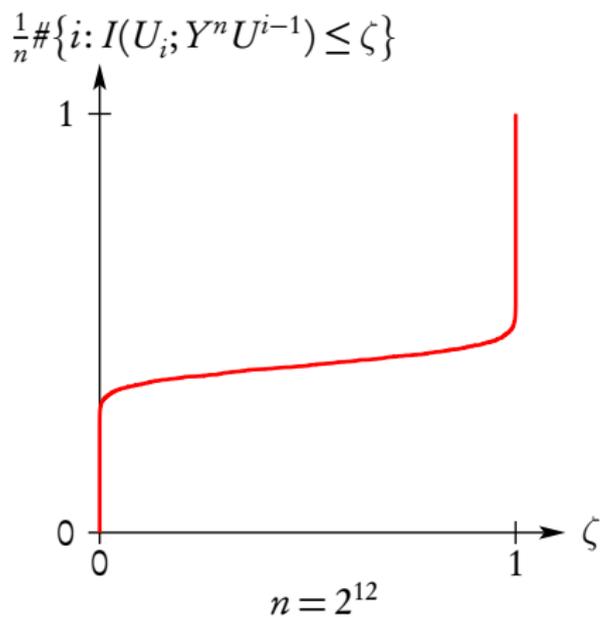
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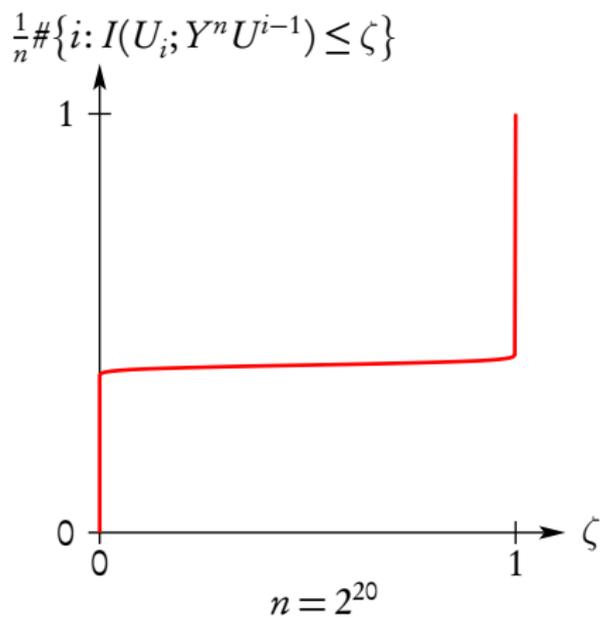
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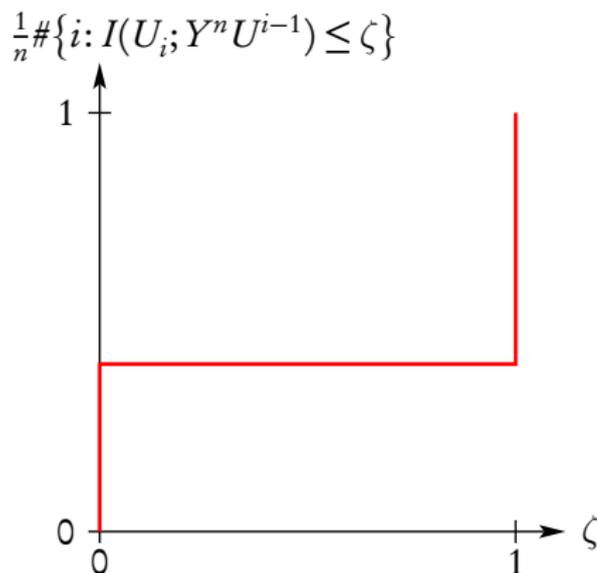
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# Polarization

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- *If* this happens, the *good* limiting synthetic channels can be used to transmit **uncoded** data bits — the inputs to the other channels can be frozen to fixed values. Since the transformation is information lossless, the fraction of good channels (i.e., data rate) must be  $I(W)$ .

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- The decoder can decode  $U_1, U_2, \dots, U_n$  successively.

# On successive decoding

The quantities  $I(U_i; Y^n U^{i-1})$  are relevant for a **genie-aided** decoder:

$$\hat{U}_1 = \phi_1(Y^n)$$

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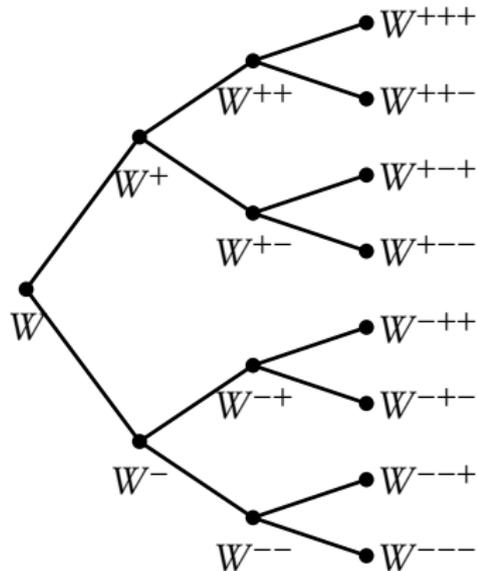
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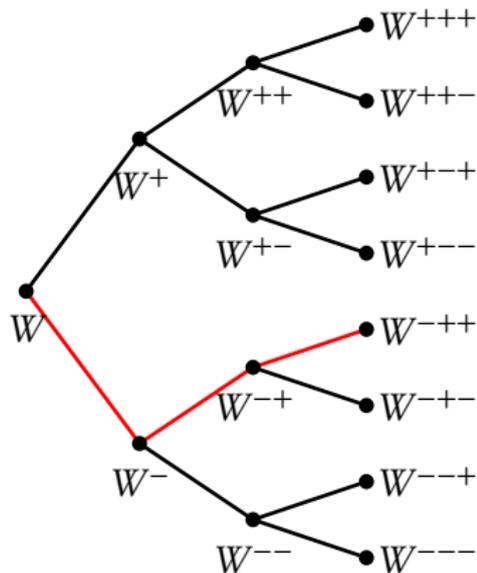
# Polarization does happen

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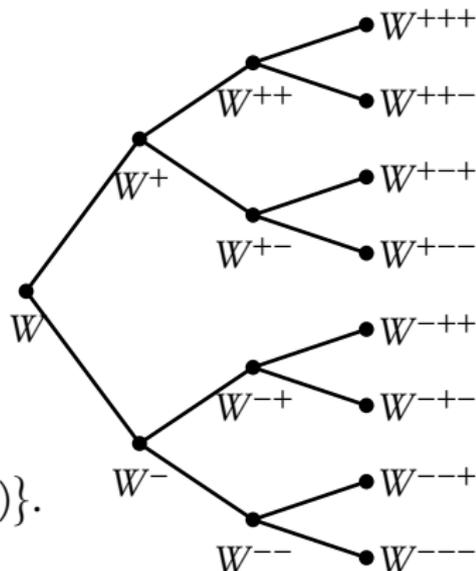
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- $W_\ell$  is uniformly distributed among  $\{W^{-----}, \dots, W^{++++}\}$ , and

$$\Pr(I(W_\ell) \in (\epsilon, 1 - \epsilon)) = \frac{1}{n} \#\{i: I(U_i; Y^n U^{i-1}) \in (\epsilon, 1 - \epsilon)\}.$$



# Polarization does happen

To show polarization all we need to study is the process  $I_\ell := I(W_\ell)$  and show that  $I_\ell \in (\epsilon, 1 - \epsilon)$  with small probability.

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- so,

$$E[I_{\ell+1} | B_1, \dots, B_\ell] = \frac{1}{2}[I(W_\ell^-) + I(W_\ell^+)] = I(W_\ell) = I_\ell$$

and we see that  $\{I_\ell\}$  is a martingale.

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- But how fast? Fast enough to arrest error propagation?
- Introduce the Bhattacharyya parameter

$$Z(W) = \sum_y \sqrt{W(y|0)W(y|1)}$$

as a companion to  $I(W)$ . Note that this is an upper bound on probability of error for uncoded transmission over  $W$ .

# Polarization speed

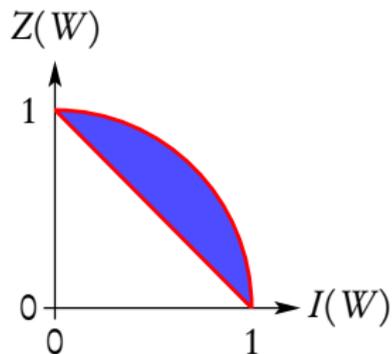
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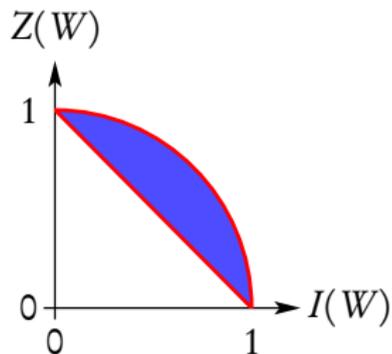
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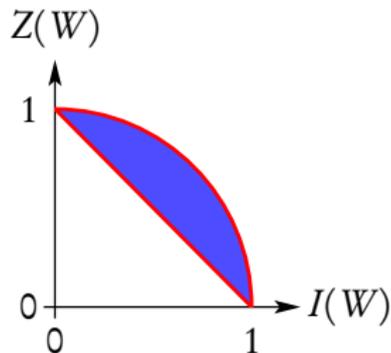
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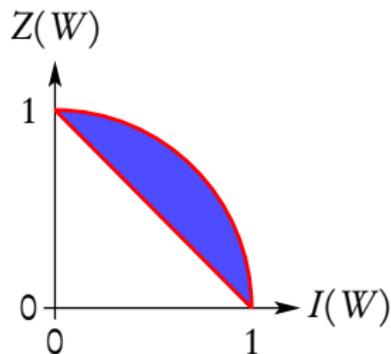
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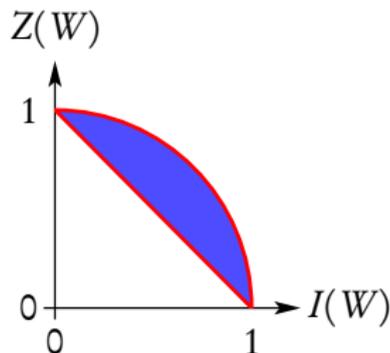
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Since  $Z(W)$  upper bounds on probability of error for uncoded transmission over  $W$ , we can choose the **good indices** on the basis of the  $Z$ 's of the synthetic channels. The sum of the  $Z$ 's of the chosen channels will upper bound the block error probability. This suggests studying the polarization speed of  $Z$ .

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- This means that for any  $\beta < 1/2$ , as long as  $R < I(W)$  the error probability of polarization codes decays to 0 faster than  $2^{-n^\beta}$ .

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- With ‘ $k$ -by- $k$ ’ constructions the error probability can be made to decay almost exponentially in blocklength.

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- For channel coding applications, the usual suspects (LDPC, Turbo, ...) easily beat the pure polar codes at meaningful block lengths. For quantization they are a lot more competitive.

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- Successive decoding is cheap ( $n \log n$ ) but too naive. More clever decoding methods should improve error probability at moderate  $n$ .
- While the original motivation was for channel coding, polar codes are good at many settings. [“Polar codes are good for everything — S. Korada”]