

Lecture 14: Polar codes

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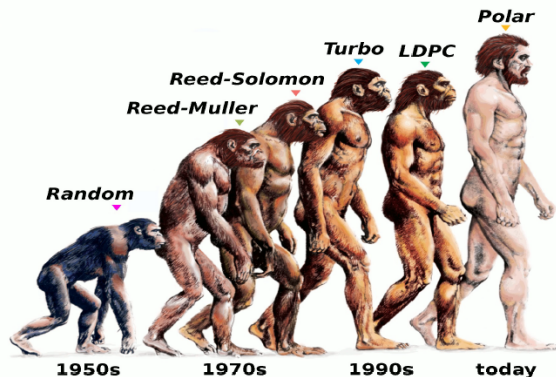
March 7, 2017

1 Polarization

2 Encoding

3 Decoding

Introduction



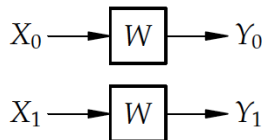
- Polar codes are the “*next big thing*” in coding theory since the advent of turbo codes and LDPC codes in the late 1990s.
- In terms of performance/complexity trade-off, the only competition to polar codes today are *spatially-coupled LDPC codes*.

1 Polarization

2 Encoding

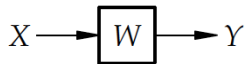
3 Decoding

A new class of error-correcting codes, invented in 2009 by Erdal Arıkan, based on the universal phenomenon of ***channel polarization***.



Two independent uses of the channel W

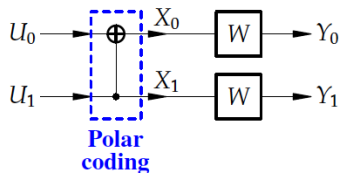
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Binary-input symmetric DMC

Polarization

A new class of error-correcting codes, invented in 2009 by Erdal Arıkan, based on the universal phenomenon of **channel polarization**.



Polar code of length $n = 2$

As the mapping is invertible, we have

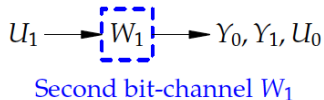
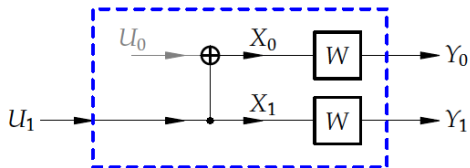
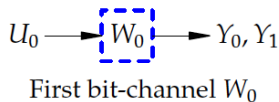
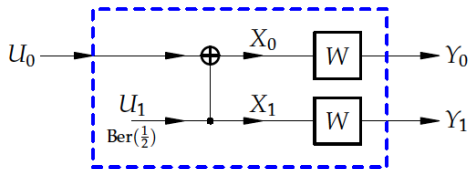
$$\begin{aligned} I(U_0, U_1; Y_0, Y_1) &= I(X_0, X_1; Y_0, Y_1) = I(X_0; Y_0) + I(X_1; Y_1) \\ &= 2I(X_1; Y_1) = 2I(W). \end{aligned}$$

The chain rule decomposition

$$\begin{aligned} I(U_0, U_1; Y_0, Y_1) &= I(U_0; Y_0, Y_1) + I(U_1; Y_0, Y_1 | U_0) \\ &= 2I(W). \end{aligned}$$

Consider “virtual” sub-channels!

“Virtual” sub-channels



“Virtual” sub-channels

Two channels are created

$$(W, W) \rightarrow (W_0, W_1)$$

W_0 is worse than W , W_1 is better than W . Will be denoted also as W^- and W^+ .

Example

Polarization is easy to analyze when W is a BEC.

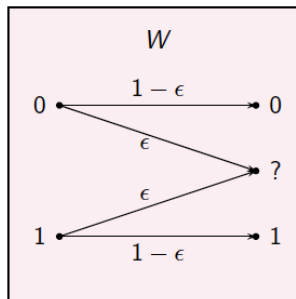
If W is a $\text{BEC}(\epsilon)$, then so are W^- and W^+ , with erasure probabilities

$$\epsilon^- \triangleq 2\epsilon - \epsilon^2$$

and

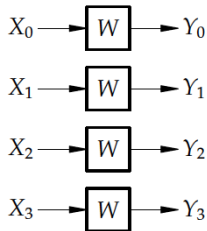
$$\epsilon^+ \triangleq \epsilon^2$$

respectively.



$$n = 4$$

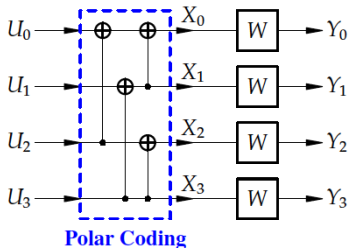
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Four independent uses of the channel W

$$n = 4$$

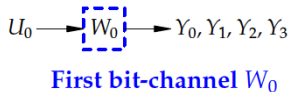
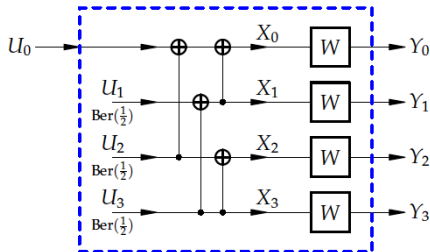
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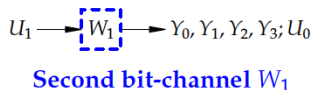
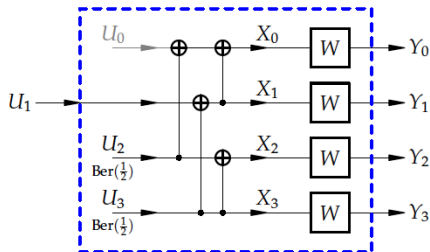
Polar code of length $n = 4$

$$n = 4$$

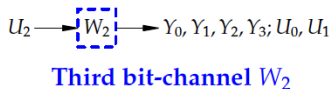
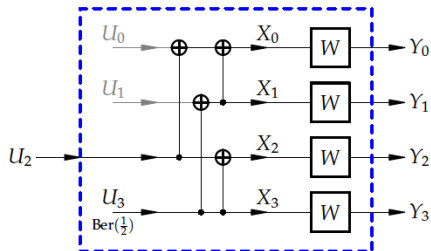
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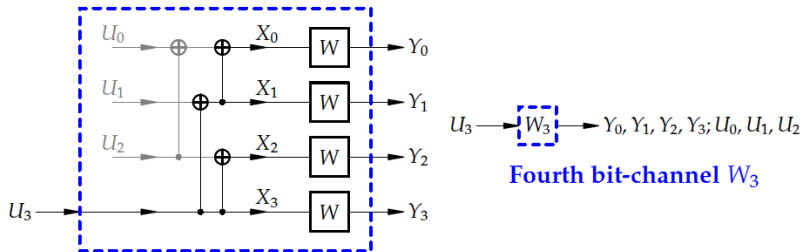
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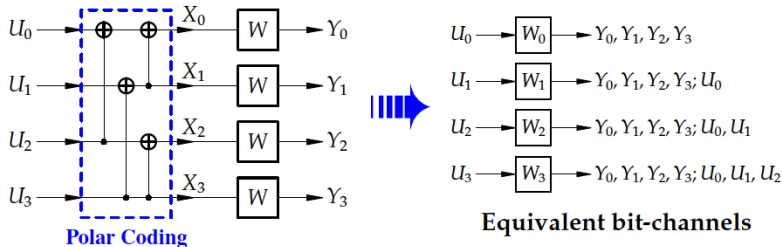


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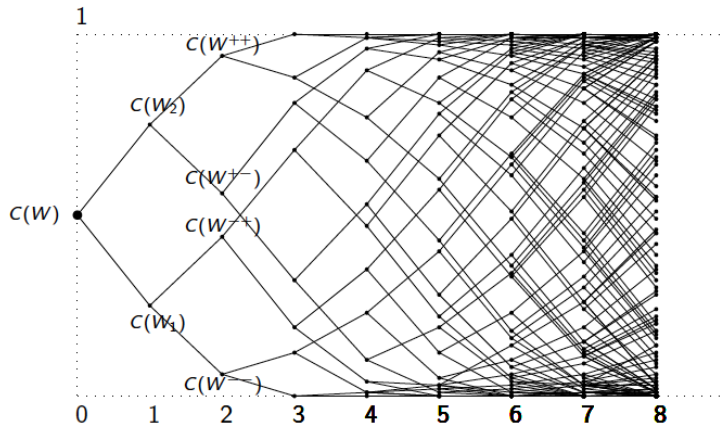


Polarization Theorem [Arıkan 2009]

As $n \rightarrow \infty$, the bit-channels polarize: they become either good (nearly noiseless) or bad (nearly useless). The fraction of good channels approaches capacity (W).

Polar codes: send information bits over the good channels, freeze the input to the bad channels to *a priori* known values (say, zeros).

General n



1 Polarization

2 Encoding

3 Decoding

Kroneker product

$$G_2 = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$$

Kronecker product

$$\mathbf{G}_2^{\otimes 2} = \begin{bmatrix} \textcolor{red}{G}_2 & 0 \\ \textcolor{blue}{G}_2 & \textcolor{green}{G}_2 \end{bmatrix}$$

Kronecker product

$$G_2^{\otimes 2} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

Kronecker product

$$G_2^{\otimes 3} = \begin{bmatrix} G_2 & 0 & 0 & 0 \\ G_2 & G_2 & 0 & 0 \\ G_2 & 0 & G_2 & 0 \\ G_2 & G_2 & G_2 & G_2 \end{bmatrix}$$

Kroneker product

$$G_2^{\otimes 3} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

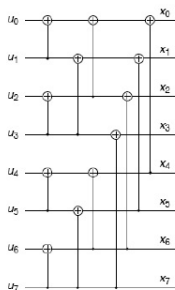
Kronecker product

length $N = 2^m$, $m \in \mathbb{N}$

generator matrix: rows of $G_2^{\otimes m}$

$$G_2^{\otimes 3} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$\bar{x} = [u_0 \ u_1 \ u_2 \ u_3 \ u_4 \ u_5 \ u_6 \ u_7] G_2^{\otimes 3}$$



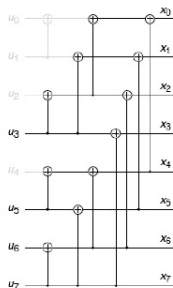
Kronecker product

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$$G_2^{\otimes 3} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$\bar{x} = [0 \ 0 \ 0 \ u_3 \ 0 \ u_5 \ u_6 \ u_7] G_2^{\otimes 3}$$



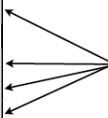
How to choose the rows?

Reed-Muller codes

length $N = 2^m$, $m \in \mathbb{N}$

generator matrix: rows of $G_2^{\otimes m}$

How to choose the rows?

$$G_2^{\otimes 3} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$


choose rows of largest weight

$$\bar{x} = [0 \ 0 \ 0 \ u_3 \ 0 \ u_5 \ u_6 \ u_7] G_2^{\otimes 3}$$

$$W = \text{BEC}(0.5), N = 8, R = 1/2$$

$I(W_i)$

0.0039

0.1211

0.1914

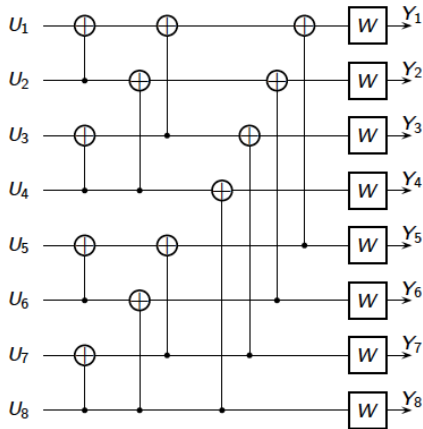
0.6836

0.3164

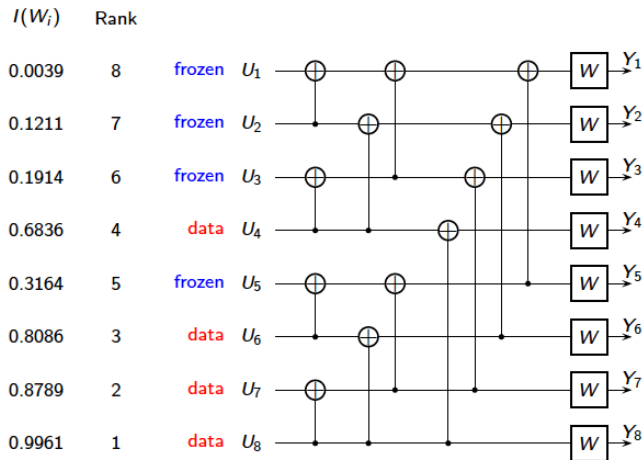
0.8086

0.8789

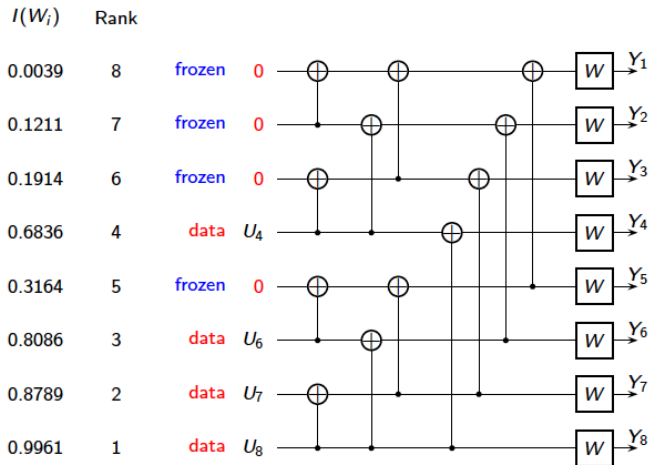
0.9961



$$W = \text{BEC}(0.5), N = 8, R = 1/2$$



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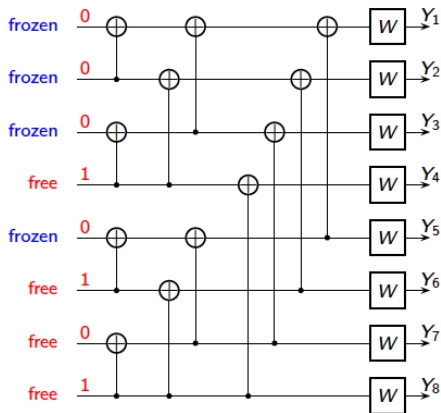
Theorem

Encoding complexity for polar coding is $\mathcal{O}(N \log N)$.

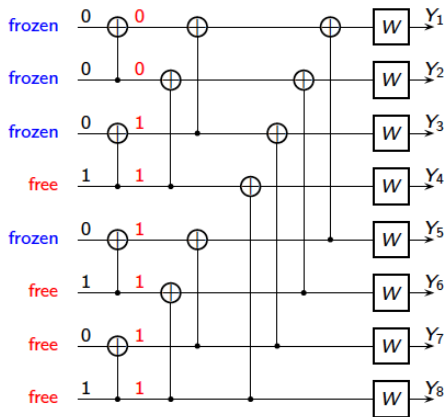
Proof:

- ▶ Polar coding transform can be represented as a graph with $N[1 + \log(N)]$ variables.
- ▶ The graph has $(1 + \log(N))$ levels with N variables at each level.
- ▶ Computation begins at the source level and can be carried out level by level.
- ▶ Space complexity $\mathcal{O}(N)$, time complexity $\mathcal{O}(N \log N)$.

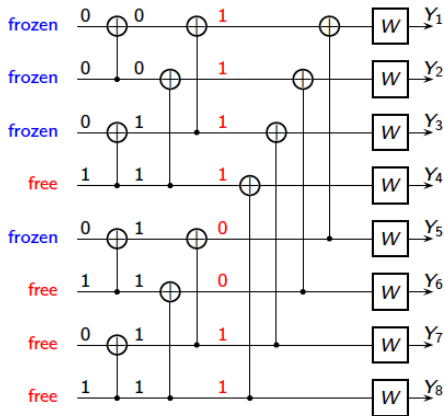
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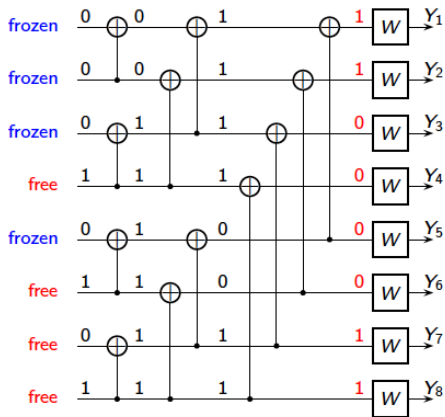
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$$W = \text{BEC}(0.5), N = 8, R = 1/2$$

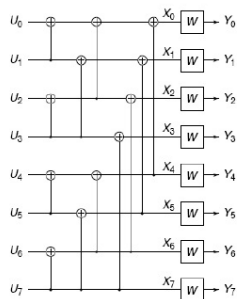


1 Polarization

2 Encoding

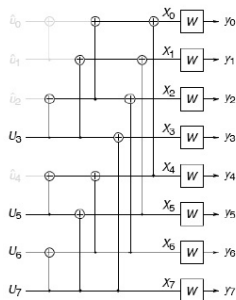
3 Decoding

Decoding



Decoding

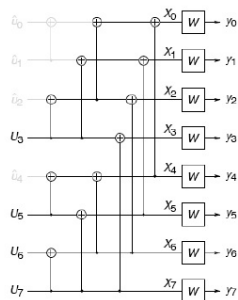
$$F = \{0, 1, 2, 4\}$$



Decoding

$$F = \{0, 1, 2, 4\}$$

From 0 till $N - 1$



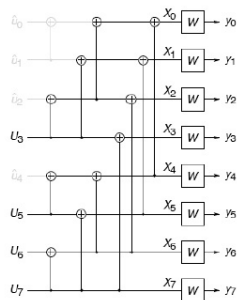
$$F = \{0, 1, 2, 4\}$$

From 0 till $N - 1$

if $i \in F$, $\hat{u}_i = 0$

if $i \in F^c$,

$$\hat{u}_i = \begin{cases} 0, & \text{if } \frac{P(0|\hat{u}_0^{i-1}, \tilde{y})}{P(1|\hat{u}_0^{i-1}, \tilde{y})} > 1 \\ 1, & \text{otherwise} \end{cases}$$



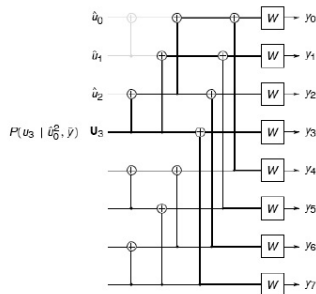
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Decoding

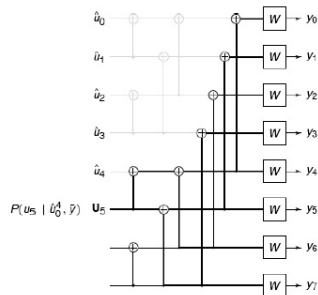
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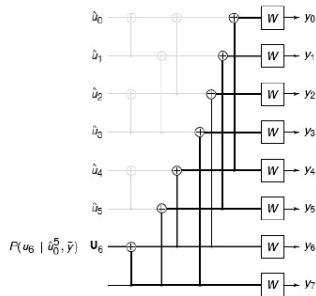
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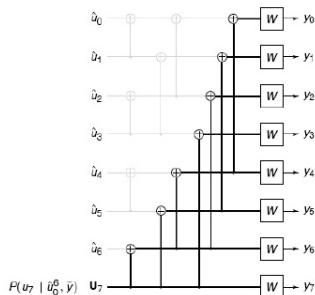
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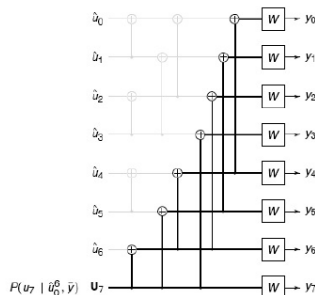
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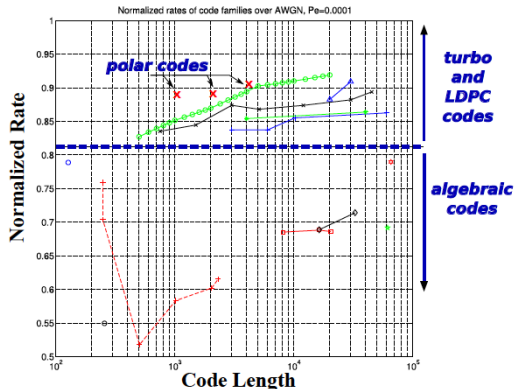
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Complexity $O(N \log N)$



Comparison to another coding schemes

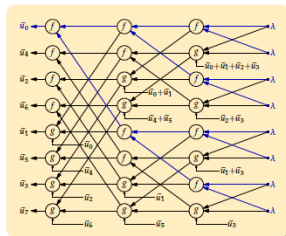


- Polar codes achieve **better performance** than the best-known LDPC codes with lower complexity, even at very short code lengths.

Summary

- Polar codes provably **achieve capacity** on general symmetric channels.
- Polar codes have **low encoding and decoding complexity** — both $O(n \log n)$.
- Polar codes **are explicit**: there is no random ensemble to choose from; they also do not suffer from an **error-floor**.
- Polar codes have **beautiful structure** that resembles the fast Fourier transform. This makes them **well suited for VLSI** implementation.

G. Sarkis, P. Giard, A. Vardy, C. Thibault, and W.J. Gross,
Fast polar decoders: Algorithm and implementation, *IEEE Journal on Selected Areas in Communications*, 32, pp. 946–957, May 2014.
- Polar codes work in **many important scenarios** other than point-to-point channel coding: wiretap channels, broadcast channels, multi-user channels, Wyner-Ziv coding, source coding, and more...



Thank you for your attention!