

# Channel Coding at Low Capacity

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## Motivation

◆ Narrowband communication: NB-IoT, eMCT [17-18]

By 2021, there will be 1.5 billion IoT devices



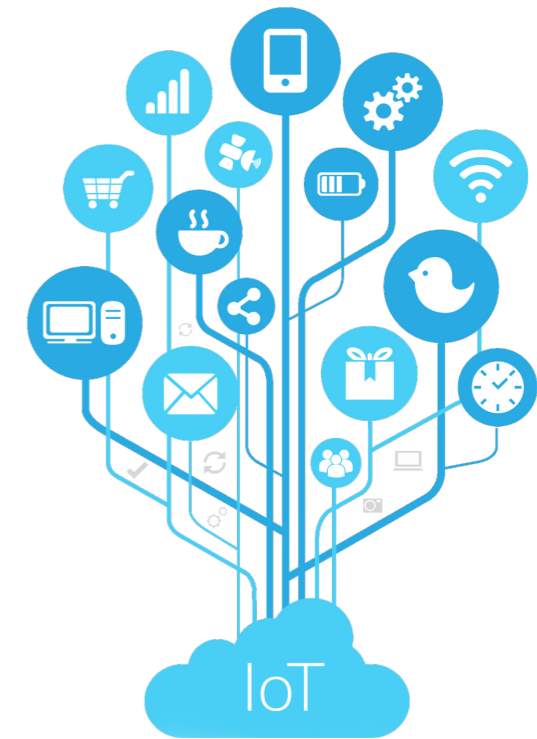
This infers a 170 dB of coupling loss



Which leads to having a -13dB of the effective SNR



This can be translated to a very small capacity of 0.03

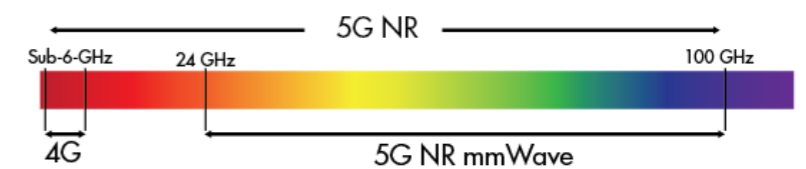
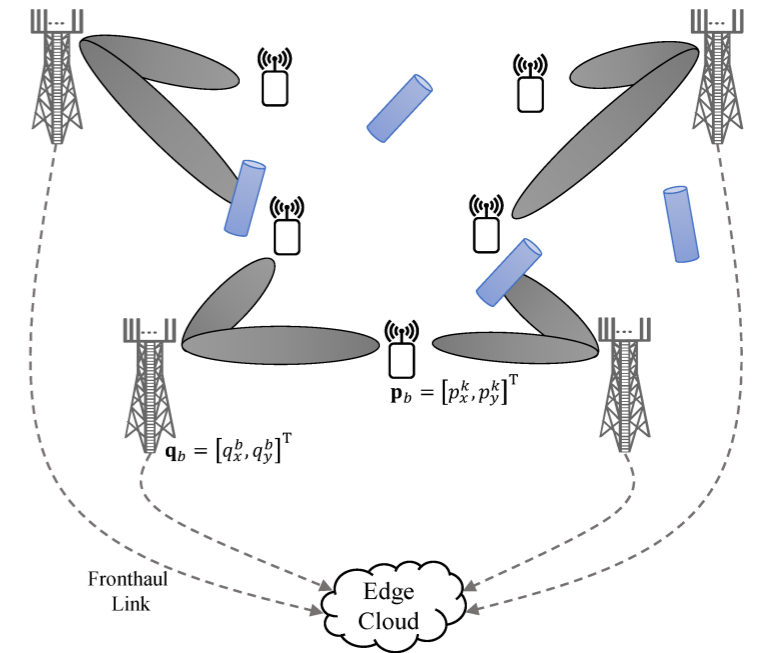


# Motivation

- ◆ Wideband communication: mmWave [17-18]

$$C = B \log\left(1 + \frac{P}{N_0 B}\right)$$

$$B \rightarrow \infty \implies SNR \rightarrow 0 \implies C \rightarrow 0$$



# Low Capacity Regime

The capacity  $C$  is small with respect to the blocklength  $n$  [2]

## ◆ Formal definition

$C < n^{s-1}$ , where  $s \in [0, 1)$  is a tuning parameter

$s$  can be specified depending on

- the application
- the channel under discussion

## ◆ Informal definition

$$\kappa := nC = o(n)$$

$$\kappa = \mathcal{O}(1) \quad \Leftrightarrow \quad C = \mathcal{O}\left(\frac{1}{n}\right) \quad \Leftrightarrow \quad s = 0$$

$$\kappa = \mathcal{O}(\sqrt{n}) \quad \Leftrightarrow \quad C = \mathcal{O}\left(\frac{1}{\sqrt{n}}\right) \quad \Leftrightarrow \quad s = \frac{1}{2}$$

# Challenges

- ◆ In the moderate-capacity regime [1,4,7]

$$\log_2 M^*(n, p_e) = nC - \sqrt{nV} Q^{-1}(p_e) + \mathcal{O}(\log_2 n)$$

This is the best that can be achieved in the moderate-capacity regime.  
(The last term depends on the type of the channel [8-II])

- ◆ In the low-capacity regime [2], however,

- using the above prediction leads to off numeral estimates since some neglected terms will become significant and also because the governing laws are different.
- Moreover, the current practical code designs are not efficient in this regime.

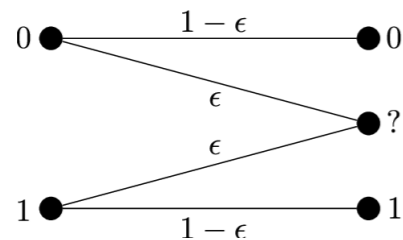
# Our Contribution

- ◆ Deriving non-asymptotic laws for BEC and BSC in the low-capacity regime
- ◆ Proposing a provably efficient practical code design for the low-capacity regime

# Binary Erasure Channel (BEC)

Let  $\epsilon$  be the erasure probability

◆ In the moderate-capacity regime: e.g.,  $\epsilon = 0.3, 0.6$

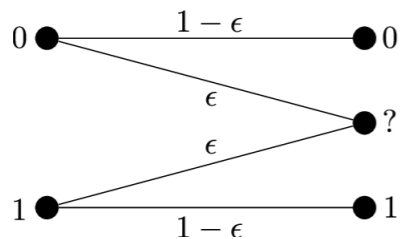


output: ...I??O?I???OO?I??OIHO??IO???III?I??.....

Gaussian Convergence Laws

$$R = n(1 - \epsilon) + \mathcal{O}(\sqrt{n})$$

◆ In the low-capacity regime: e.g.,  $\epsilon = 0.95, 0.99$



output: ...?????I?????O????????????I??????I??.....

Rare information  $\rightarrow$  Poisson Convergence Laws [I6]

$$P\{R < r\} = \mathcal{P}_{n(1-\epsilon)}(r)$$

# Binary Erasure Channel (BEC)

$$M^*(n, p_e) = \max \{M \mid \exists (M, p_e)\text{-code for } W^n\}$$

where  $n$  is the blocklength and  $p_e$  is the average probability of error

The raw achievability bound (RCU) for BEC [1,5,6]

$$p_e \leq \sum_{r=0}^n \binom{n}{r} \epsilon^{n-r} (1 - \epsilon)^r 2^{-[r - \log_2(M-1)]^+}$$

The raw converse bound for BEC [1,5,6]

$$p_e \geq \sum_{r < \log_2 M} \binom{n}{r} \epsilon^{n-r} (1 - \epsilon)^r \left(1 - \frac{2^r}{M}\right)$$



# Binary Erasure Channel (BEC)

The raw achievability bound (RCU)

The raw converse bound

Using **Gaussian** convergence laws for the **moderate-capacity** regime, due to [I]

Using **Poisson** convergence laws for the **low-capacity** regime + a finer analysis

$$\log_2 M^*(n, p_e) = nC - \sqrt{nV} Q^{-1}(p_e) + \mathcal{O}(1)$$

$$C = 1 - \epsilon$$

$$V = \epsilon(1 - \epsilon)$$

Our result on the next page

## Binary Erasure Channel (BEC)

**Theorem.** Consider transmission over  $\text{BEC}(\epsilon)$  in low-capacity regime and let  $\kappa = n(1 - \epsilon)$ . Then,

$$M_1 \leq M^*(n, p_e) \leq M_2,$$

where  $M_1$  is the solution of

$$\mathfrak{P}_1(M_1) + \alpha \sqrt{\mathfrak{P}_1(M_1)} - p_e = 0, \quad (1)$$

and  $M_2$  is the solution of

$$\mathfrak{P}_2(M_2) - \alpha \sqrt{\mathfrak{P}_2(M_2)} - \alpha \sqrt{\mathcal{P}_\kappa(\log_2 M_2)} - p_e = 0, \quad (2)$$

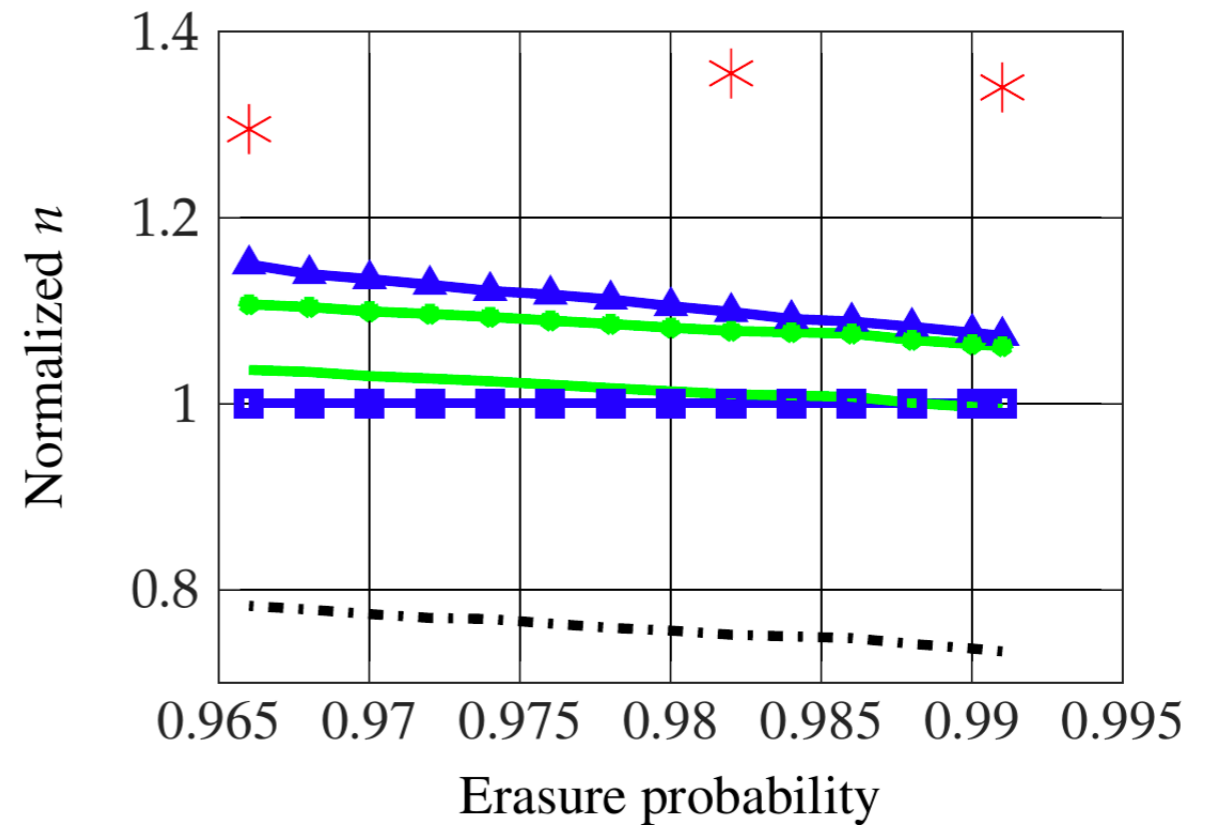
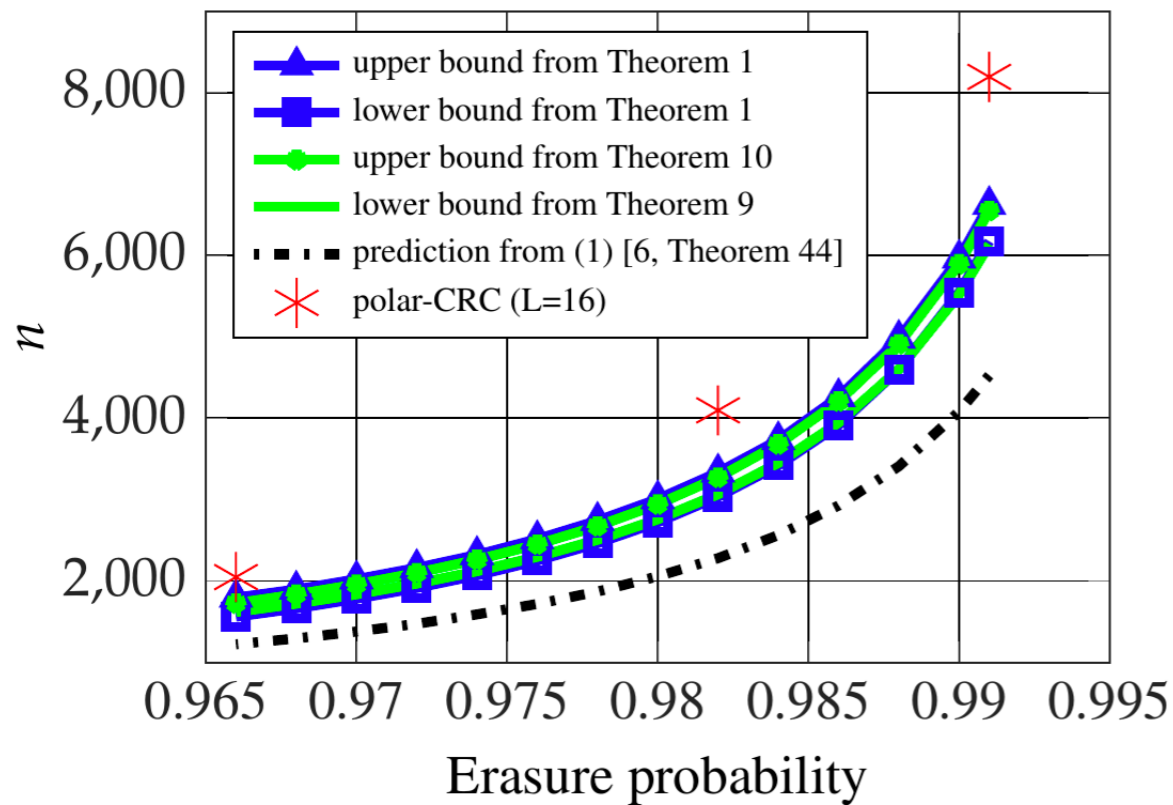
and

$$\mathfrak{P}_1(M_1) = \mathcal{P}_\kappa(\log_2 M_1) + M_1 e^{-\kappa/2} (1 - \mathcal{P}_{\kappa/2}(\log_2 M_1)),$$

$$\mathfrak{P}_2(M_2) = \mathcal{P}_\kappa(\log_2 M_2) - \frac{e^\kappa}{M_2} \mathcal{P}_{2\kappa}(\log_2 M_2),$$

$$\alpha = \frac{\sqrt{2}}{\epsilon^{3/2}} \left( 1 + 2\sqrt{\frac{3}{\epsilon\kappa}} \right) (\sqrt{e} - 1) (1 - \epsilon).$$

# Binary Erasure Channel (BEC)



At each  $\epsilon$  we plotted the optimal  $n$  such that  $M^*(n, 10^{-2}) = 2^{40}$

**Green:** The raw upper and lower bounds for the reality

**Blue:** Our prediction of the bounds

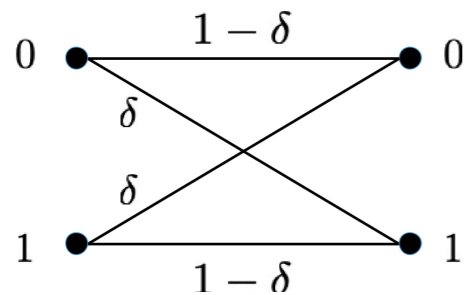
**Black:** The prediction from [1] (A single prediction for both upper and lower bounds)

The polar code is concatenated with cyclic redundancy check (CRC) code of length 6, and is decoded with the list-SC algorithm [3] with list size  $L=16$ .

# Binary Symmetric Channel (BSC)

Let  $\delta$  be the crossover probability

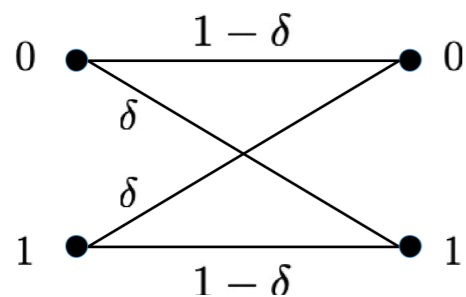
◆ In the moderate-capacity regime: e.g.,  $\delta = 0.2, 0.9$



output: ...IOIOOOIHIOOOIHIOIOIHIOIOIOOOIHIOIHIO.....

## Gaussian Convergence Laws

◆ In the low-capacity regime: e.g.,  $\delta = 0.46, 0.52$



output: ...OOIOIOIHIOOOIOIHIOOOIHIOOHIOIHIO.....

## Again Gaussian Convergence Laws

but needs a finer analysis since some neglected terms are significant now

# Binary Symmetric Channel (BSC)

The raw achievability bound (RCU) for BSC [1,5,6]

$$p_e \leq \sum_{r=0}^n \binom{n}{r} \delta^r (1 - \delta)^{n-r} \min \left\{ 1, (M - 1) S_n^r \right\}$$

$$S_n^r = \sum_{s=0}^r \binom{n}{s} 2^{-n}$$

The raw converse bound for BSC [1,5,6]

$$M \leq \frac{1}{\beta_{1-p_e}^n}$$

where  $\beta_\alpha^n$  for a real  $\alpha \in [0, 1]$  is defined below based on values of  $\beta_\ell$  where  $\ell$  is an integer:

$$\beta_\alpha^n = (1 - \lambda) \beta_L + \lambda \beta_{L+1},$$

$$\beta_\ell = \sum_{r=0}^{\ell} \binom{n}{r} 2^{-n},$$

such that  $\lambda \in [0, 1)$  and integer  $L$  satisfy the following:

$$\alpha = (1 - \lambda) \alpha_L + \lambda \alpha_{L+1},$$

$$\alpha_\ell = \sum_{r=0}^{\ell-1} \binom{n}{r} \delta^r (1 - \delta)^{n-r}.$$

# Binary Symmetric Channel (BSC)

The raw achievability bound (RCU)

The raw converse bound

for the **moderate-capacity** regime,  
due to [I]

for the **low-capacity** regime  
+ a finer analysis

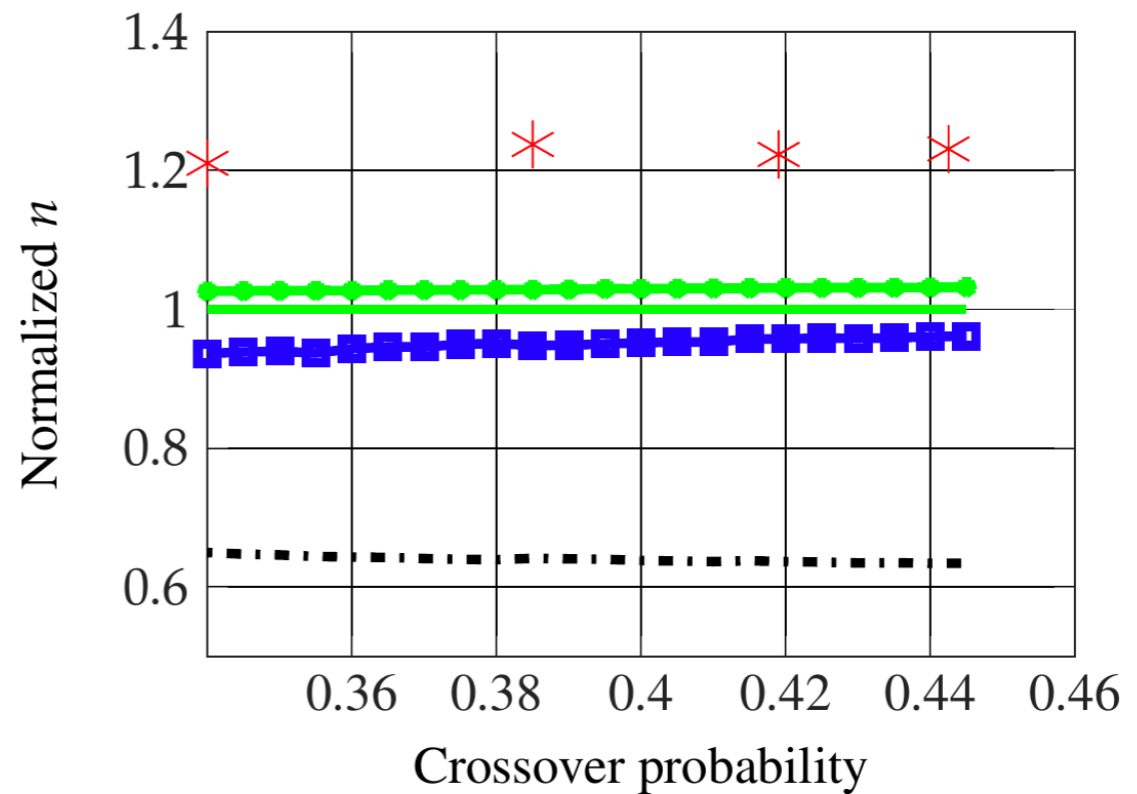
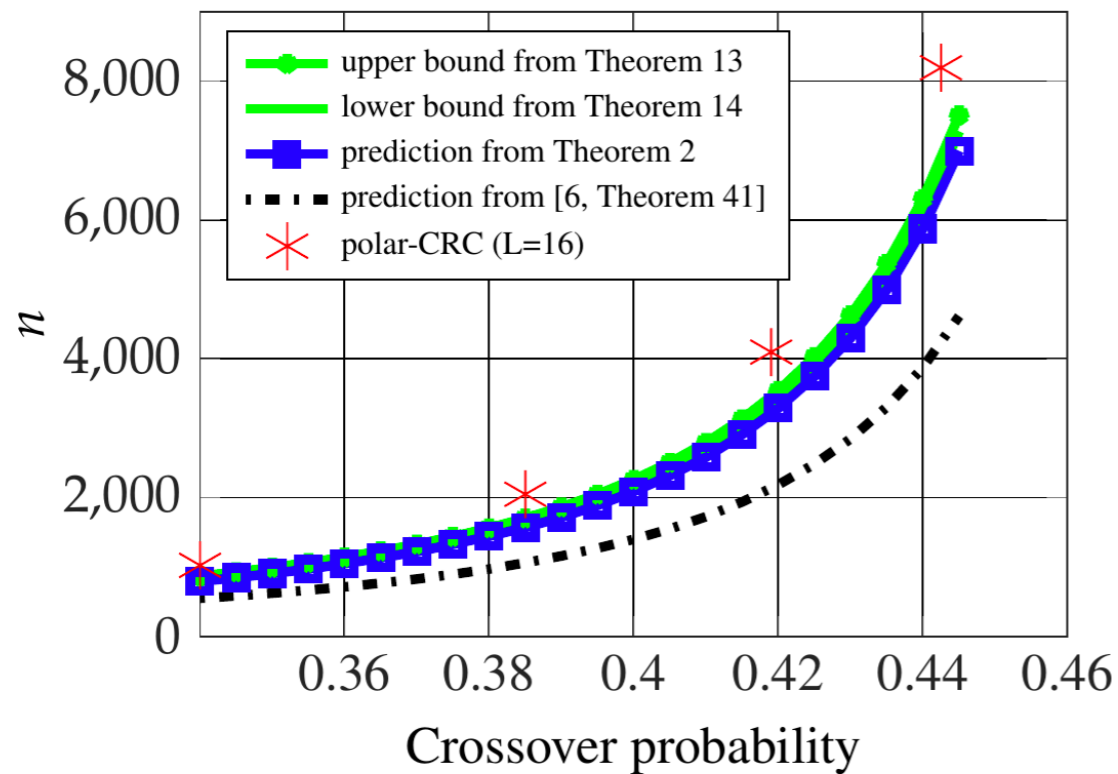
$$\log_2 M^*(n, p_e) = nC - \sqrt{nV} Q^{-1}(p_e) + \frac{1}{2} \log_2 n + \mathcal{O}(1)$$

$$C = 1 - h_2(\delta)$$

$$V = \delta(1 - \delta) \log_2^2((1 - \delta)/\delta)$$

Theorem.  $\log_2 M^*(n, p_e) = \kappa - 2\sqrt{\frac{2\delta(1 - \delta)}{\ln 2}} \cdot \sqrt{\kappa} Q^{-1}(p_e) + \frac{1}{2} \log_2 \kappa - \log_2 p_e + \mathcal{O}(\log \log \kappa)$   
 $\kappa = n(1 - h_2(\delta))$

# Binary Symmetric Channel (BSC)



At each  $\delta$  we plotted the optimal  $n$  such that  $M^*(n, 10^{-2}) = 2^{40}$

**Green:** The raw upper and lower bounds for the reality

**Blue:** Our prediction of the bounds (A single prediction for both upper and lower bounds)

**Black:** The prediction from [1] (A single prediction for both upper and lower bounds)

The polar code is concatenated with cyclic redundancy check (CRC) code of length 6, and is decoded with the list-SC algorithm [3] with list size  $L=16$ .

# Practical Code Design

In summary:

- ◆ Current standards use repetition + some off-the-shelf moderate-rate codes like Turbo codes/ LDPC.
- ◆ Using iterative codes + repetition results in mediocre performances.
- ◆ **Polar codes** implicitly apply the optimal repetition length.

The next slides argue these results in detail.



# Practical Code Design

Consider the largest (optimal) repetition size  $r_\beta$  which satisfies

$$\frac{n}{r_\beta} C(W^{r_\beta}) \geq \beta n C(W)$$

**Theorem.** If  $W = \text{BEC}(\epsilon)$ , then for the largest repetition size,  $r_\beta$ , we have

$$\frac{n(1-\epsilon)\ell}{2\left(1-\frac{\beta}{\ell}\right)} \cdot \left(\frac{\beta}{\ell}\right)^2 \leq \frac{n}{r_\beta} \leq \frac{n(1-\epsilon)\ell}{2\left(1-\frac{\beta}{\ell}\right)},$$

where  $\ell = -\frac{\ln \epsilon}{1-\epsilon}$ . Equivalently, assuming  $\kappa = n(1-\epsilon)$ , we have

$$\frac{\kappa}{2(1-\beta)} \cdot \beta^2(1 + \mathcal{O}(1-\epsilon)) \leq \frac{n}{r_\beta} \leq \frac{\kappa}{2(1-\beta)}(1 + \mathcal{O}(1-\epsilon)).$$

# Practical Code Design

The previous result about BEC leads to the following theorem

**Theorem.** Among all BMS channels with the same capacity, BEC has the largest repetition length  $r_\beta$ . Hence, for any BMS channel with capacity  $C$  and  $\kappa = nC$ , we have

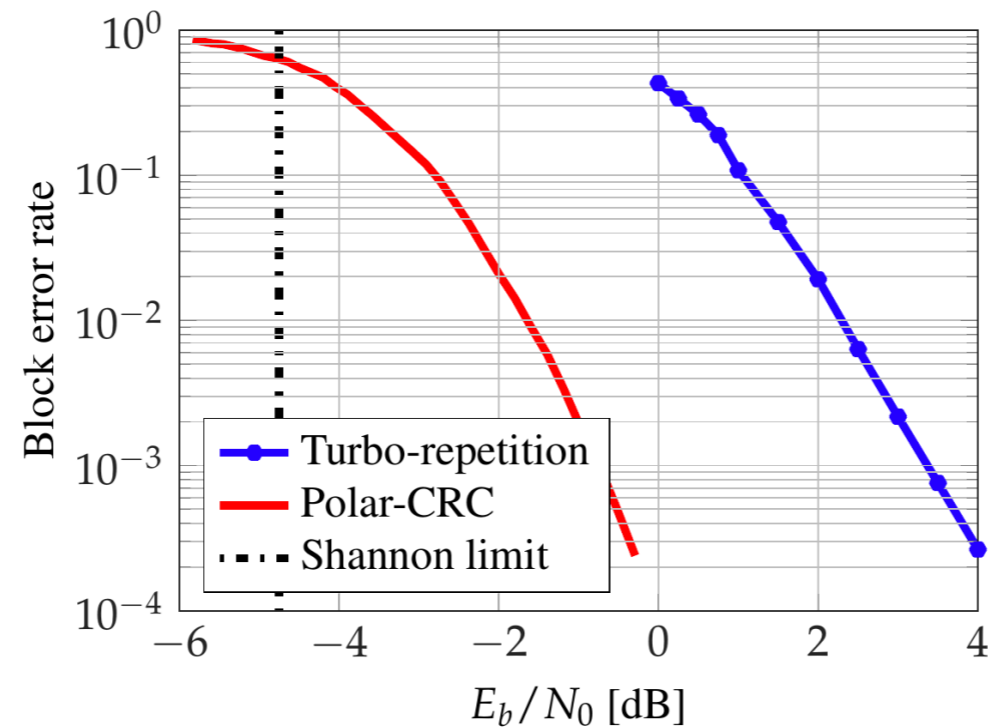
$$\frac{n}{r_\beta} \geq \frac{\kappa}{2(1-\beta)} \beta^2 (1 + \mathcal{O}(1-C)).$$

**Remark.** we can conclude that for any BMS channel with low capacity, in order to have the total rate loss of order  $\mathcal{O}(1)$ , the repetition size should be at most  $\mathcal{O}(n/\kappa^2)$ .

Now the following theorem shows that polar codes automatically perform a repetition coding which matches the optimal repetition size  $\mathcal{O}(n/\kappa^2)$ .

**Theorem.** Consider using a polar code of length  $n = 2^m$  for transmission over a BMS channel  $W$ . Let  $m_0 = \log_2(4\kappa^2)$  where  $\kappa = nC(W)$ . Then any synthetic channel  $W_n^{(i)}$  whose Bhattacharyya value is less than  $\frac{1}{2}$  has at least  $m_0$  plus operations in the beginning. As a result, the polar code constructed for  $W$  is equivalent to the concatenation of a polar code of length (at most)  $2^{m_0}$  followed by  $2^{m-m_0}$  repetitions.

# Practical Code Design



- Comparison for low-capacity BAWGN. The number of information bits is  $k=40$ .
- The polar-CRC has length 8192, is constructed using 6 CRC bits, and is decoded using the SC-list decoder with  $L=16$ .
- The Turbo-repetition has an underlying (120,40) Turbo code which is repeated 68 times (total length = 8160) and is decoded with 6 iterations.
- The Shannon limit for this setting is -4.75dB.

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**Thank you!**