

Blocklength Scaling of Polar Codes

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Problem Definition

- ▶ $W(\cdot | \cdot)$ – DMC.
- ▶ $C = I(W)$ – symmetric capacity.
- ▶ Goal: Communicate at rate R with error probability $P_e \leq P_e^0$.
- ▶ Capacity achieving family of codes: For any $R < C$, can find code with rate R and blocklength N , such that $P_e \leq P_e^0$.
- ▶ How does N scale with respect to $C - R$?
- ▶ Without complexity considerations: $N = \beta / (C - R)^2$ (best possible and is achievable).
- ▶ What about finite length scaling of computationally efficient capacity achieving codes?

Blocklength Scaling of Binary Polar Codes

- ▶ Polar codes [Arikan, 2009] are capacity achieving.
- ▶ Computational complexity is $O(N \log N)$.
- ▶ Blocklength scales polynomially: $N = \frac{\beta}{(C-R)^\mu}$
 [Guruswami and Xia, 2013], [Hassani et al., 2014]. How small can we set μ ?
 - ▶ $3.55 \leq \mu \leq 6$ [Hassani et al., 2014].
 - ▶ $\mu \leq 5.7$ [Goldin and Burshtein, 2014].
 - ▶ $\mu \leq 4.7$ [Mondelli et al., 2015].
- ▶ Similar scaling of N in lossy source coding, w.r.t. $R(D) - R$ [Goldin and Burshtein, 2014] and various problems in multiuser information theory.

How can we generalize / improve results?

- ▶ General polarization kernels, or nonbinary polar codes.
- ▶ Recent result for q -ary polar codes when q is prime: N scales polynomially with respect to $\frac{1}{C-R}$: $N = \frac{\beta}{(C-R)^\mu}$
[Guruswami and Velingker, 2014].
- ▶ However, in the proof, μ is very large. Can we do better?
- ▶ We show that for $q = 3$ much lower values of μ can be obtained
[Goldin and Burshtein, 2015].
- ▶ The technique can be applied to other values of prime q .

Polarization

Proposed in [Arikan, 2009]

- ▶ Blocklength $N = 2^n$
- ▶ Generator matrix G_N , size $N \times N$
- ▶ Message vector $\mathbf{u} = u_1^N$, $\mathbf{x} = x_1^N = \mathbf{u}G_N$
- ▶ B-DMC channel $W : \mathcal{X} \rightarrow \mathcal{Y}$, $\mathcal{X} = \{0, 1\}$
- ▶ Channel output $\mathbf{y} = y_1^N$
- ▶ Probability distribution: $P(\mathbf{u}, \mathbf{x}, \mathbf{y}) = \frac{1}{2^N} \mathbb{1}_{\{\mathbf{x}=\mathbf{u}G_N\}} \prod_{i=1}^N W(y_i | x_i)$
- ▶ For $i = 1, 2, \dots, N$, define the N sub-channels

$$W_N^{(i)}(\mathbf{y}, u_1^{i-1} | u_i) \triangleq P(\mathbf{y}, u_1^{i-1} | u_i) = \frac{1}{2^{N-1}} \sum_{u_{i+1}^N} P(\mathbf{y} | \mathbf{u})$$

- ▶ *Polarization*: Typically, either $I(W_N^{(i)}) \approx 1$ or $I(W_N^{(i)}) \approx 0$.

Polar codes, Encoding

- ▶ Code rate $R < I(W)$.
- ▶ Let $Z(W) \triangleq \sum_{y \in \mathcal{Y}} \sqrt{W(y|0)W(y|1)}$.
- ▶ The frozen set F is the set of $N(1 - R)$ sub-channels with highest $Z(W_N^{(i)})$.

Algorithm (Encoding)

- ▶ If $i \in F$, fix to frozen \mathbf{u}_F .
- ▶ If $i \in F^c$ use it for information.
- ▶ Transmit $\mathbf{x} = \mathbf{u}G_N$.

Polar codes, Decoding

Algorithm (Decoding)

For $i = 1, 2, \dots, N$:

1. If $i \in F$, $\hat{u}_i = u_i$
2. If $i \in F^c$, $\hat{u}_i = \begin{cases} 0 & \text{if } L_N^{(i)} > 1 \\ 1 & \text{if } L_N^{(i)} \leq 1 \end{cases}$ where $L_N^{(i)} = \frac{W_N^{(i)}(\mathbf{y}, \hat{\mathbf{u}}_1^{i-1} | u_i=0)}{W_N^{(i)}(\mathbf{y}, \hat{\mathbf{u}}_1^{i-1} | u_i=1)}$

- ▶ For $R < I(W)$, error probability, P_e , satisfies [Arikan and Telatar, 2009]:

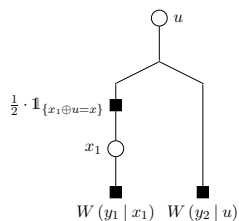
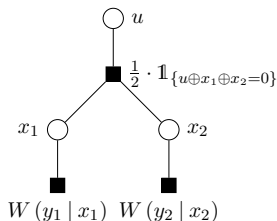
$$P_e = O\left(2^{-N^\beta}\right), \quad \text{for any } \beta < 1/2$$

- ▶ Encoding and decoding complexity $O(N \log N)$.

Analysis of Polarization

Sub-channels can be described using the following random process:

- ▶ $B_1, B_2 \dots$ i.i.d $\Pr\{B_n = 0\} = \Pr\{B_n = 1\} = 1/2$
- ▶ $W_0 = W, W_{n+1} = \begin{cases} W_n^-, & \text{if } B_{n+1} = 0 \\ W_n^+, & \text{if } B_{n+1} = 1. \end{cases}$
 - ▶ $W^-(y_1, y_2 | u) \triangleq (W \otimes W)(y_1, y_2 | u) \triangleq \frac{1}{2} \sum_x W(y_1 | u \oplus x) W(y_2 | x)$
 - ▶ $W^+(y_1, y_2, x | u) \triangleq (W \otimes W)(y_1, y_2, x | u) \triangleq \frac{1}{2} W(y_1 | x \oplus u) W(y_2 | u)$



Analysis of Polarization (CONT'D)

- ▶ W_n uniformly distributed over $\{W_N^{(i)}\}_{i=1}^N$.
- ▶ Hence, for $Z_n = Z(W_n)$, $I_n = I(W_n)$,

$$P[Z_n \in (a, b)] = \left| \left\{ i : Z(W_N^{(i)}) \in (a, b) \right\} \right| / N$$

$$P[I_n \in (a, b)] = \left| \left\{ i : I(W_N^{(i)}) \in (a, b) \right\} \right| / N$$

- ▶ It was shown [Arikan, 2009], for any fixed small $\delta > 0$,
 - ▶ $\lim_{n \rightarrow \infty} \Pr(Z_n \leq \delta) = I(W)$
 - ▶ $\lim_{n \rightarrow \infty} \Pr(Z_n \geq 1 - \delta) = 1 - I(W)$
 - ▶ $\lim_{n \rightarrow \infty} \Pr(I_n \leq \delta) = 1 - I(W)$
 - ▶ $\lim_{n \rightarrow \infty} \Pr(I_n \geq 1 - \delta) = I(W)$

How can finite length scaling be derived?

Following [Hassani et al., 2014] and the variations in [Goldin and Burshtein, 2014]:

- ▶ It is known that

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

$$Z(W)\sqrt{2 - Z^2(W)} \leq Z(W^-) \leq 2Z(W) - Z^2(W)$$

- ▶ For some $f_0(z) > 0$, $z \in (0, 1)$, $f_0(0) = f_0(1) = 0$, define $f_k(z)$ recursively:

$$f_k(z) \triangleq \sup_{y \in [z\sqrt{2-z^2}, z(2-z)]} \frac{f_{k-1}(z^2) + f_{k-1}(y)}{2}$$

- ▶ Also define $L_k(z) \triangleq f_k(z)/f_0(z)$, $L_k \triangleq \sup_{z \in (0,1)} L_k(z)$.
- ▶ It can be shown that $E[f_0(Z_n)] \leq A \cdot (\sqrt[k]{L_k})^n \cdot f_0[Z(W)]$ for constant A .

How can finite length scaling be shown? (CONT'D)

- ▶ Using appropriately chosen $f_0(z)$ it can now be shown:

$$P(Z_n \in (\delta, 1 - \delta)) \leq \frac{A}{\delta} \left(\sqrt[k]{L_k} \right)^n \leq \frac{A}{\delta} 2^{-\rho n}$$

for constant A and $\rho = 0.2127$.

- ▶ Proceed by showing, given m_0 , for constant \tilde{A} , that

$$P(\omega \in \Omega : Z_n(\omega) \notin (\delta, 1 - \delta) \forall n \geq m_0) \geq 1 - \frac{\tilde{A}}{\delta} 2^{-\rho m_0}$$

$$P(\omega \in \Omega : Z_n(\omega) \leq \delta \forall n \geq m_0) \geq I(W) - \frac{\tilde{A}}{\delta} 2^{-\rho m_0}$$

How can finite length scaling be shown? (CONT'D)

- ▶ Following [Arıkan, 2009] it can now be shown that for

$$R \leq I(W) - \left(1 + \frac{A}{\delta}\right) \cdot 2^{-\alpha n}$$

we have

$$P_e = O(N^{-a})$$

where $a > 0$ for $\alpha = 1/(1 + 1/\rho) = 5.702^{-1}$.

- ▶ This proves the following scaling result:

Theorem

For $P_e \leq P_e^0$, sufficient to set $N = \beta / (I(W) - R)^{5.702}$.

Outline of analysis of q -ary polarization

- ▶ Instead of $Z(W_n)$ use $I(W_n)$.
- ▶ Given q -ary input channel W , $W^- = W \boxtimes W$ and $W^+ = W \circledast W$ obtain a bound

$$I(W) - I(W^-) \geq \epsilon_l [I(W)]$$

for some $\epsilon_l [I(W)]$.

- ▶ For some $f_0(x) > 0$, $x \in (0, 1)$, $f_0(0) = f_0(1) = 0$, define $f_k(x)$, for $k = 1, 2, \dots$, recursively

$$f_k(x) \triangleq \sup_{\epsilon_l(x) \leq \epsilon \leq \epsilon_h(x)} \frac{f_{k-1}(x + \epsilon) + f_{k-1}(x - \epsilon)}{2}$$

for $\epsilon_h(x) \triangleq \min(x, 1 - x)$.

- ▶ The rest of the analysis is very similar to the binary case.

Outline of analysis of q -ary polarization (CONT'D)

- ▶ In particular $L_k(x) \triangleq f_k(x)/f_0(x)$, $L_k \triangleq \sup_{x \in (0,1)} L_k(x)$.
- ▶ Hence

$$\begin{aligned}
 & \mathbb{E}[f_k(I_{n+1})] \\
 &= \mathbb{E}\left[\frac{f_k(I_n^+) + f_k(I_n^-)}{2}\right] \\
 &\leq \mathbb{E}\left[\sup_{\epsilon_l(x) \leq \epsilon \leq \epsilon_h(x)} \frac{f_k(I_n + \epsilon) + f_k(I_n - \epsilon)}{2}\right] \\
 &\leq \mathbb{E}[f_{k+1}(I_n)]
 \end{aligned}$$

- ▶ Hence $\mathbb{E}[f_0(I_n)] \leq \mathbb{E}[f_k(I_{n-k})] \leq L_k \mathbb{E}[f_0(I_{n-k})]$.
- ▶ Hence $\mathbb{E}[f_0(I_n)] \leq A \cdot (\sqrt[k]{L_k})^n \cdot f_0[I(W)]$ for constant A .
- ▶ Rest is almost identical to the binary case when using Z_n .

The main difficulty

- ▶ In the binary case $q = 2$, a tight bound $\epsilon_l [I(W)]$ such that $I(W) - I(W^-) \geq \epsilon_l [I(W)]$ is well known, e.g. [Richardson and Urbanke, 2008].
- ▶ This is not the case for $q > 2$.
- ▶ We show how good bounds can be obtained numerically.

Our approach to obtain $\epsilon_l(x)$

- ▶ Following notation in [Karzand and Telatar, 2010], given q -ary channel $W(y|x)$

$$W(y) \triangleq (1/q) \sum_{x=0}^{q-1} W(y|x)$$

$$\mathbf{v}(y) \triangleq [v_0(y), v_1(y), \dots, v_{q-1}(y)]^T$$

$$v_x(y) \triangleq \frac{W(y|x)}{qW(y)}, \quad \sum_{x=0}^{q-1} v_x(y) = 1$$

- ▶ Then: $I(W) = \sum_y W(y) [1 - H[\mathbf{v}(y)]] = \sum_G \hat{W}(G) \cdot G$ where $\hat{W}(G) \triangleq \sum_{y: H[\mathbf{v}(y)]=1-G} W(y)$

Our approach to obtain $\epsilon_l(x)$ (CONT'D)

- ▶ Given two channels, W_a and W_b , let $W_{a\boxtimes b} \triangleq W_a \boxtimes W_b$, i.e.

$$W_{a\boxtimes b}(y_1, y_2 | u) \triangleq \frac{1}{q} \sum_{u'=0}^{q-1} W_b(y_2 | u') W_a(y_1 | u + u')$$

- ▶ Hence $W_{a\boxtimes b}(y_1, y_2) = W_a(y_1) W_b(y_2)$ and
[Karzand and Telatar, 2010]

$$\mathbf{v}_{a\boxtimes b}(y_1, y_2) = \mathbf{v}_b(y_2) \star \mathbf{v}_a(y_1)$$

where \star denotes q -circular cross-correlation.

- ▶ Also define

$$g(G_1, G_2) \triangleq 1 - \min_{\substack{H[\mathbf{v}_a(y_1)] = 1 - G_1 \\ H[\mathbf{v}_b(y_2)] = 1 - G_2}} H[\mathbf{v}_b(y_2) \star \mathbf{v}_a(y_1)]$$

Our approach to obtain $\epsilon_l(x)$ (CONT'D)

$$\begin{aligned}
 I(W_{a \boxtimes b}) &= \sum_{y_1, y_2} W_{a \boxtimes b}(y_1, y_2) \{1 - H[\mathbf{v}_{a \boxtimes b}(y_1, y_2)]\} \\
 &\leq \sum_{G_1, G_2} \sum_{\substack{y_1: H[\mathbf{v}_a(y_1)] = 1 - G_1 \\ y_2: H[\mathbf{v}_b(y_2)] = 1 - G_2}} W_a(y_1) W_b(y_2) g(G_1, G_2) \\
 &= \sum_{G_1, G_2} \hat{W}_a(G_1) \hat{W}_b(G_2) g(G_1, G_2)
 \end{aligned}$$

If $g(G_1, G_2)$ concave (separately!) in G_1, G_2 (otherwise replace by concave upper bound)

$$I(W_{a \boxtimes b}) \leq g \left[\sum_{G_1} \hat{W}_a(G_1) G_1, \sum_{G_2} \hat{W}_a(G_2) G_2 \right] = g[I(W_a), I(W_b)]$$

Our approach to obtain $\epsilon_l(x)$ (CONT'D)

- ▶ In our case $W^- = W \boxtimes W$. Hence $I(W^-) \leq g[I(W), I(W)]$.
- ▶ Hence $I(W) - I(W^-) \geq I(W) - g[I(W), I(W)] \triangleq \epsilon_l[I(W)]$.
- ▶ Recall

$$g(G_1, G_2) \triangleq 1 - \min_{\substack{H[\mathbf{v}_a(y_1)] = 1 - G_1 \\ H[\mathbf{v}_b(y_2)] = 1 - G_2}} H[\mathbf{v}_b(y_2) \star \mathbf{v}_a(y_1)]$$

- ▶ At least for $q = 3$, a QSC channel provides an excellent approximation to the solution!
- ▶ A QSC with error prob. p :

$$W(y | x) = \begin{cases} 1 - p & y = x \\ p/(q - 1) & y \neq x \end{cases}$$

Properties of $g(G_1, G_2)$

$$g(G_1, G_2) \triangleq 1 - \min_{\substack{H[\mathbf{v}_a(y_1)] = 1 - G_1 \\ H[\mathbf{v}_b(y_2)] = 1 - G_2}} H[\mathbf{v}_b(y_2) \star \mathbf{v}_a(y_1)]$$

Lemma

If W_a and W_b are QSC, then $W_{a \boxtimes b}$ is QSC, and $I(W_{a \boxtimes b}) = g_{QSC}[I(W_a), I(W_b)]$.

Lemma

Using QSC channels W_a and W_b yields extreme point in Lagrangian of definition of $g(G_1, G_2)$, $\forall G_1, G_2 > 0$.

Properties of $g(G_1, G_2)$ (CONT'D)

Lemma

$$g(G_1, G_2) = 1 - \min_{\substack{H[\mathbf{v}_a(y_1)] \geq 1-G_1 \\ H[\mathbf{v}_b(y_2)] \geq 1-G_2}} H[\mathbf{v}_b(y_2) \star \mathbf{v}_a(y_1)]$$

Lemma

Define $f(\mathbf{u}) \triangleq \min_{H(\mathbf{v}) \geq 1-G} H(\mathbf{u} \star \mathbf{v})$. Then, $f(\mathbf{u})$ is concave.

$g(G_1, G_2)$ can be computed efficiently using algorithms for concave minimization over convex region.

Properties of $g(G_1, G_2)$ (CONT'D)

Lemma

1. $g(G_1, G_2) = g(G_2, G_1)$
2. $g(x_1, y_1) \leq g(x_2, y_2)$ for $x_1 \leq x_2$ and $y_1 \leq y_2$.
3. $g(1, G_2) = G_2$
4. $g(G_1, G_2) \leq \min(G_1, G_2)$.
5. $\lim_{x \rightarrow 1} \frac{\partial g(x, G_2)}{\partial x} = 0$

Lemma

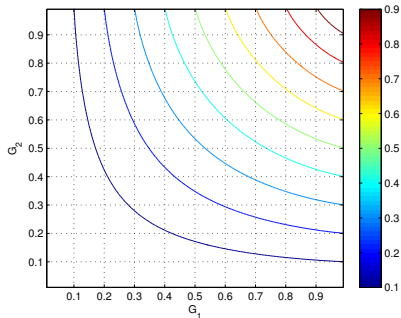
For sufficiently small G_1, G_2 and $q = 3$, $g(G_1, G_2) = \ln 3 \cdot G_1 G_2$.

Lemma

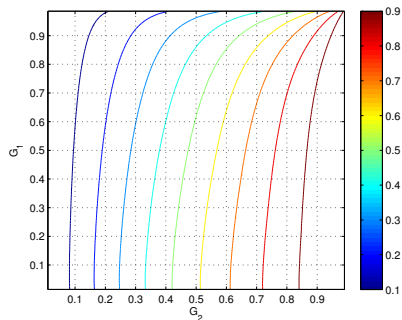
For G_1, G_2 sufficiently close to 1, and $q = 3$, $g(G_1, G_2) = G_1 + G_2 - 1$

Numerical Results

$g(G_1, G_2)$ for $q = 3$

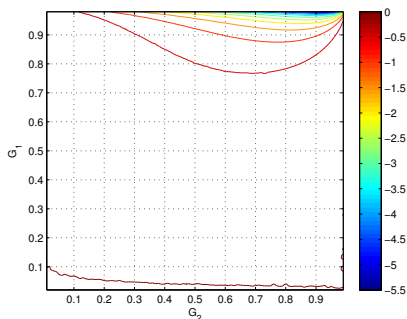


$\frac{\partial g(G_1, G_2)}{\partial G_1}$ for $q = 3$



Numerical Results (CONT'D)

$$\frac{\partial^2 g(G_1, G_2)}{\partial G_1^2} \text{ for } q = 3$$



We can find a concave upper bound on $g(G_1, G_2)$.

A concave upper bound on $g(G_1, G_2)$

- ▶ For a given G_2 , concave hull of $g(G_1, G_2)$ is obtained by passing a tangent line:

$$\max_{x \in [G_1, 1]} \frac{G_1}{x} g(x, G_2)$$

- ▶ In order to obtain upper bound on $g(G_1, G_2)$, concave in G_1 and G_2 (separately):

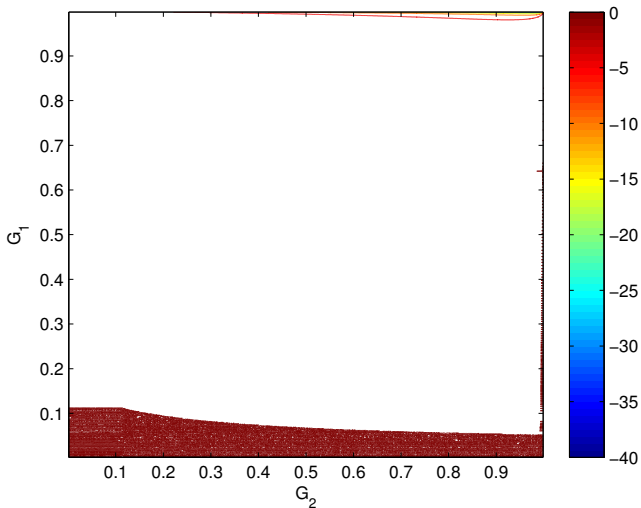
$$g^*(G_1, G_2) = \max_{x_1 \in [G_1, 1], x_2 \in [G_2, 1]} \frac{G_1 G_2}{x_1 x_2} g(x_1, x_2)$$

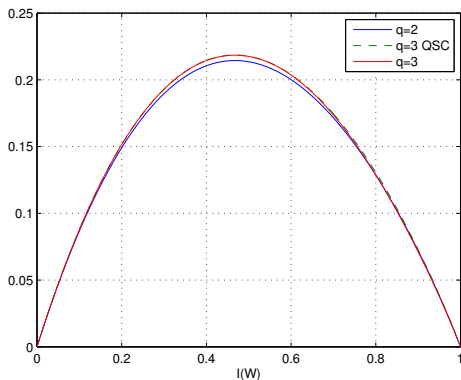
- ▶ We can also obtain closed form concave upper bound on $g(G_1, G_2)$ given by

$$g_{QSC}^*(G_1, G_2) + 0.0104[G_1(1 - G_2) + G_2(1 - G_1)]$$

However this solution produces a slightly worse bound on the scaling.

$$\frac{\partial^2 g_{QSC}^*(G_1, G_2)}{\partial G_1^2} \text{ for } q = 3$$



Lower bound on $I(W) - I(W^-)$ 

Using this bound (with $g^*(G_1, G_2)$), can be shown that scaling of N is

$$N = \frac{\beta}{(I(W) - R)^{6.504}} \text{ (or better), } \beta = \beta(P_e^0).$$

Conclusion

- ▶ The blocklength of polar codes scales polynomially with respect to the inverse gap between code rate and capacity.
- ▶ For binary and ternary polar codes this polynomial has low degree.
- ▶ The numerical technique presented may also work for other nonbinary polar codes.
- ▶ May be interesting to examine the dependence of the scaling parameter in the bound w.r.t. the alphabet size (q). Does it decrease w.r.t. q ?

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