

Polar Coding for Processes with Memory

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¹Intel ²Technion

- ▶ Well known: polarization occurs for a memoryless process
- ▶ Our setting: a process with memory
- ▶ Mild assumption: (ψ -mixing, $\psi_0 < \infty$)
- ▶ New: both weak and fast polarization occur under mild assumption
- ▶ New: example of a stationary periodic process that does not polarize

Process:

- ▶ $(X_j, Y_j, S_j)_{j=-\infty}^{\infty}$
- ▶ Polarization applied to X_j : $U_1^N = X_1^N G_N$
- ▶ Y_j channel output/side information
- ▶ S_j process state (usually hidden)

Entropy:

$$\mathcal{H}_{X|Y} = \lim_{N \rightarrow \infty} \frac{1}{N} H(X_1^N | Y_1^N)$$

Theorem (Weak polarization)

If process is ψ mixing with $\psi_0 < \infty$, then for all $\epsilon > 0$

$$\lim_{N \rightarrow \infty} \frac{1}{N} |\{i : H(U_i | U_1^{i-1} Y_1^N) > 1 - \epsilon\}| = \mathcal{H}_{X|Y},$$
$$\lim_{N \rightarrow \infty} \frac{1}{N} |\{i : H(U_i | U_1^{i-1} Y_1^N) < \epsilon\}| = 1 - \mathcal{H}_{X|Y}.$$

Theorem (Fast polarization)

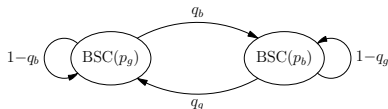
If process is ψ mixing with $\psi_0 < \infty$, then for all $\beta < 1/2$

$$\lim_{N \rightarrow \infty} \frac{1}{N} |\{i : Z(U_i | U_1^{i-1} Y_1^N) < 2^{-N^\beta}\}| = 1 - \mathcal{H}_{X|Y}.$$

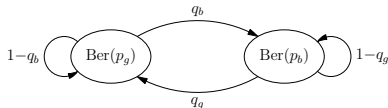
Missing: Fast polarization to entropy 1...

Even so: Above theorems \implies

- ▶ polar coding transmission scheme for the Gilbert-Elliott channel



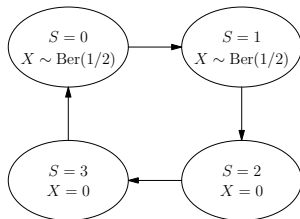
- ▶ polar coding lossless compression scheme for sources with memory



See also: R. Wang, J. Honda, H. Yamamoto, R. Liu, and Y. Hou, "Construction of polar codes for channels with memory," in *Proc. IEEE Inform. Theory Workshop (ITW'2015)*, Jeju Island, Korea, 2015, pp. 187–191.

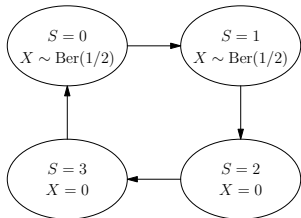
Theorem (Periodic processes may not polarize)

The stationary periodic Markov process



does not polarize. Indeed, for all $\frac{5N}{8} < i \leq \frac{6N}{8}$,

$$\left| H(U_i | U_1^{i-1}) - \frac{1}{2} \right| \leq \epsilon_N, \quad \lim_{N \rightarrow \infty} \epsilon_N = 0.$$



Lemma

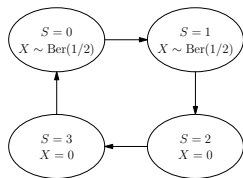
Consider the stationary Markov process depicted in the figure. Then, for $N \geq 8$, the following holds.

For all $\frac{5N}{8} < i \leq \frac{6N}{8}$ we have that

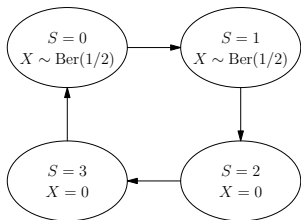
$$H(U_i | U_1^{i-1}, S_1 = s_1) = \begin{cases} 0 & \text{if } s_1 \in \{1, 3\} \\ 1 & \text{if } s_1 \in \{0, 2\} \end{cases}$$

$$\implies H(U_i | U_1^{i-1}, S_1) = \frac{1}{2}.$$

	(U_2, U_4)	(U_1, U_3, U_5)
$S_1 = 0$	$U_4 = 0$	
$S_1 = 1$	i.i.d.	$U_5 = U_3$
$S_1 = 2$	$U_4 = U_2$	
$S_1 = 3$	i.i.d.	$U_5 = U_3 + U_1$



- ▶ Table: distribution of U_1^5 for $N = 8$ and the four possible initial states
- ▶ First column: differentiate between $S_1 = 0, S_1 = 2, S_1 \in \{1, 3\}$
- ▶ Second column: differentiate between $S_1 = 1$ and $S_1 = 3$



- ▶ Counter-examples for other periods p ?
- ▶ Specifically, is it important that $p|2$?

A process $T_j = (X_j, Y_j, S_j)$ is ψ -mixing if there is a sequence

$$\psi_0, \psi_1, \dots, \quad \lim \psi_k = 1,$$

such that

$$\Pr(A \cap B) \leq \psi_k \Pr(A) \Pr(B)$$

for all $A \in \sigma(T_{-\infty}^0)$ and $B \in \sigma(T_{k+1}^\infty)$.

Graphically:

$$\cdots T_{-2} T_{-1} T_0 T_1 T_2 \cdots T_{k-1} T_k T_{k+1} T_{k+2} T_{k+3} \cdots$$

i.i.d./aperiodic Markov/aperiodic hidden Markov $\implies \psi_0 < \infty$.

▶ Let $N = 2^n$ and $1 \leq i \leq N$.

▶ Notation:

$$U_1^N = X_1^N G_N$$

$$V_1^N = X_{N+1}^{2N} G_N$$

$$Q_i = Y_1^N U_1^{i-1}$$

$$R_i = Y_{N+1}^{2N} V_1^{i-1}$$

▶ Notation, for independent blocks:

▶ Let $\hat{X}_1^{2N}, \hat{Y}_1^{2N}$ be distributed as $P_{X_1^N Y_1^N} \cdot P_{X_{N+1}^{2N} Y_{N+1}^{2N}}$

▶ Define the corresponding variables $\hat{U}_i, \hat{V}_i, \hat{Q}_i, \hat{R}_i$ as above

▶ **Bhattacharyya**: for U and Q, define

$$Z(U|Q) = \sum_q \sqrt{P_{U,Q}(0, q) \cdot P_{U,Q}(1, q)}.$$

Proof of fast polarization:

$$\begin{aligned} & Z(U_i + V_i | Q_i, R_i) \\ &= \sum_{q,r} \sqrt{P_{U_i+V_i, Q_i, R_i}(0, q, r) \cdot P_{U_i+V_i, Q_i, R_i}(1, q, r)} \\ &\leq \sum_{q,r} \sqrt{\psi_0 P_{\hat{U}_i+\hat{V}_i, \hat{Q}_i, \hat{R}_i}(0, q, r) \cdot \psi_0 P_{\hat{U}_i+\hat{V}_i, \hat{Q}_i, \hat{R}_i}(1, q, r)} \\ &= \psi_0 \cdot Z(\hat{U}_i + \hat{V}_i | \hat{Q}_i, \hat{R}_i) \\ &\leq \psi_0 \cdot 2Z(\hat{U}_i | \hat{Q}_i) \\ &= \psi_0 \cdot 2Z(U_i | Q_i) \end{aligned}$$

In a similar manner, we show

$$Z(V_i | U_i + V_i, Q_i, R_i) \leq \psi_0 \cdot Z(U_i | Q_i)^2$$

Now, apply Arıkan and Telatar ISIT 2009, assuming weak polarization

Proof of **weak polarization**:

Recall our notation

$$U_1^N = X_1^N G_N$$

$$V_1^N = X_{N+1}^{2N} G_N$$

$$Q_i = Y_1^N U_1^{i-1}$$

$$R_i = Y_{N+1}^{2N} V_1^{i-1}$$

Lemma: If $\psi_0 < \infty$, then for any $\epsilon > 0$, the fraction of indices i for which

$$I(U_i; R_i | Q_i) < \epsilon$$

$$I(V_i; Q_i | R_i) < \epsilon$$

$$I(U_i; V_i | Q_i, R_i) < \epsilon$$

approaches 1 as $N \rightarrow \infty$.

Proof:

$$\begin{aligned}\log(\psi_0) &\geq E \left[\log \frac{p_{X_1^{2N} Y_1^{2N}}}{p_{X_1^N Y_1^N} \cdot p_{X_{N+1}^{2N} Y_{N+1}^{2N}}} \right] \\ &= I(X_1^N Y_1^N; X_{N+1}^{2N} Y_{N+1}^{2N}) \\ &= I(U_1^N Y_1^N; V_1^N Y_{N+1}^{2N}) \\ &\geq I(U_1^N; V_1^N Y_{N+1}^{2N} | Y_1^N) \\ &= \sum_{i=1}^N I(U_i; V_1^N Y_{N+1}^{2N} | Y_1^N U_1^{i-1}) \\ &= \sum_{i=1}^N I(U_i; V_{i+1}^N, V_i, R_i | Q_i)\end{aligned}$$

- ▶ At most $\sqrt{\log(\psi_0)N}$ terms inside the sum are at most $\sqrt{\log(\psi_0)/N}$
- ▶ The i th term is greater than both $I(U_i; R_i | Q_i)$ and $(U_i; V_i | Q_i, R_i)$

Lemma: Let (X_i, Y_i) be stationary and ψ -mixing. For all $\xi > 0$, there exists N_0 and $\delta(\xi) > 0$ such that for all $N > N_0$ and all $\{0, 1\}$ -valued random variables $A = f(X_1^N, Y_1^N)$ and $B = f(X_{N+1}^{2N}, Y_{N+1}^{2N})$

$$p_A(0) \in (\xi, 1 - \xi) \quad \text{implies} \quad p_{AB}(0, 1) > \delta(\xi).$$

Proof: Define the random variable $C = f(X_{2N+1}^{3N}, Y_{2N+1}^{3N})$. We have

$$\begin{aligned} 2p_{AB}(0, 1) &= p_{AB}(0, 1) + p_{BC}(0, 1) \\ &\geq p_{ABC}(0, 1, 1) + p_{ABC}(0, 0, 1) \\ &= p_{AC}(0, 1) \\ &= p_A(0) - p_{AC}(0, 0) \\ &\geq p_A(0)(1 - \psi_N p_C(0)) \\ &= p_A(0)(1 - \psi_N p_A(0)) \end{aligned}$$

- ▶ The first and last equalities are due to stationarity
- ▶ Since $p_A(0) \in (\xi, 1 - \xi)$ and $\psi_N \rightarrow 1$, there exists N_0 such that the last term is away from 0 for all $N > N_0$.

- ▶ The above two lemmas are the essence of the proof
- ▶ A proof for the case of finite memory was given in the Ph.D. thesis of Şaşıoğlu
- ▶ Current proof more general, and easier to follow (there are similarities)