

An Upgrading Algorithm with Optimal Power Law

Or Ordentlich¹ Ido Tal²

¹Hebrew University ²Technion

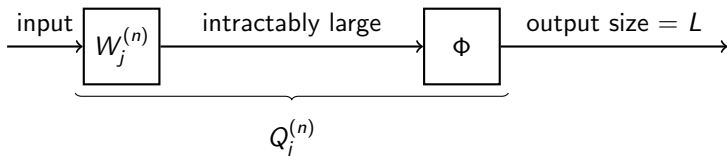
Big picture first

In this talk:

- ▶ An upgrading algorithm for channels with **non-binary** input
- ▶ **Optimal** power law
- ▶ Achieved by **reduction** to the binary-input case
- ▶ Important for constructing polar codes

Constructing vanilla polar codes

- ▶ Underlying channel: a binary-input **symmetric** and **memoryless** channel $W : \mathcal{X} \rightarrow \mathcal{Y}$, where $\mathcal{X} = \{0, 1\}$
- ▶ Derive $N = 2^n$ **synthetic channels** $W_j^{(n)} : \mathcal{X} \rightarrow \mathcal{Y}^N \times \mathcal{X}^{j-1}$, where $1 \leq j \leq N$.
- ▶ Constructing a vanilla polar code \equiv finding which synthetic channels $W_j^{(n)}$ are '**almost noiseless**'
- ▶ Problem: output alphabet $\mathcal{Y}^N \times \mathcal{X}^{j-1}$ is **intractably large**
- ▶ Solution:
 - ▶ Replace $W_j^{(n)}$ with $Q_j^{(n)}$ having output alphabet size L
 - ▶ Have $Q_j^{(n)}$ be (stochastically) **degraded** with respect to $W_j^{(n)}$



- ▶ $Q_j^{(n)}$ almost noiseless $\implies W_j^{(n)}$ almost noiseless

Constructing **vanilla** polar codes

- ▶ We write $Q \leq W$ if Q is **degraded** with respect to W
- ▶ Alternatively, we write $W \geq Q$ and say that W is **upgraded** with respect to Q

- ▶ Previous slide:

$$Q_j^{(n)} \leq W_j^{(n)}$$

- ▶ We can also approximate $W_j^{(n)}$ “from above” by an **upgraded** channel $R_j^{(n)}$ having output alphabet size at most L .
- ▶ Sandwich property:

$$Q_j^{(n)} \leq W_j^{(n)} \leq R_j^{(n)}$$

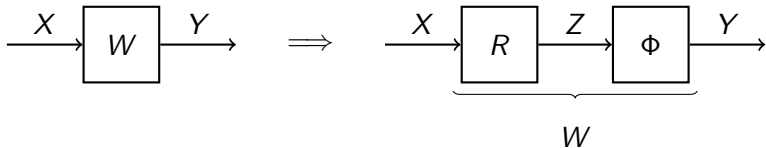
- ▶ In **vanilla** setting, $R_j^{(n)}$ has secondary importance. . .

Constructing generalized polar codes

- ▶ Polar codes have been generalized **beyond** vanilla setting
 - ▶ Asymmetric channels (with asymmetric input distribution)
 - ▶ Wiretap channels
 - ▶ Channels with memory (input distribution can have memory as well)
- ▶ In all these settings upgrading is **as important as** degrading for constructing the code
- ▶ For settings with memory, the “effective input alphabet” is **non-binary**

Problem statement

- ▶ Given: **joint** distribution of channel and input $P_{X,Y}(x,y)$
 - ▶ $x \in \mathcal{X}$, the input alphabet and $y \in \mathcal{Y}$, the output alphabet
 - ▶ $P_{X,Y}(x,y) = \underbrace{P_X(x)}_{\text{input distribution}} \cdot \underbrace{P_{Y|X}(y|x)}_{\text{channel}}$
- ▶ Find: $P_{X,Z,Y}^*(x,z,y)$ such that
 - ▶ Marginalization: $\sum_z P_{X,Z,Y}^*(x,z,y) = P_{X,Y}(x,y)$
 - ▶ Upgrading: $X - Z - Y$ is a Markov chain
 - ▶ Tractable output alphabet size: $z \in \mathcal{Z}$ and $|\mathcal{Z}| \leq L$



- ▶ **Figure of merit:**

$$H(X|Y) - H(X|Z) = I(X; Z) - I(X; Y)$$

should be 'small'

Power law

- ▶ Previous results:

- ▶ Recall: output alphabet size of upgraded channel $|\mathcal{Z}| \leq L$
- ▶ There exists a 'hard to upgrade' joint distribution $P(X, Y)$:

$$H(X|Y) - H(X|Z) = \Omega(L^{-2/(|\mathcal{X}|-1)})$$

- ▶ For **binary input**, $|\mathcal{X}| = 2$, and **any** $P_{X,Y}$, there **exists** an upgrading algorithm such that

$$H(X|Y) - H(X|Z) = O(L^{-2}) = O(L^{-2/(|\mathcal{X}|-1)})$$

- ▶ **New** result:

- ▶ Also for **non-binary** input, we can upgrade **any** $P_{X,Y}$ and achieve

$$H(X|Y) - H(X|Z) = O(L^{-2/(|\mathcal{X}|-1)})$$

- ▶ Main idea: use binary-input as a **black-box** (reduction)

One-hot representation

- ▶ Denote $q = |\mathcal{X}|$. Assume

$$\mathcal{X} = \{1, 2, \dots, q\}$$

- ▶ For $x \in \mathcal{X}$, define

$$g(x) = (x_1, x_2, \dots, x_{q-1}),$$

the **one-hot** representation:

$$g(1) = (1, 0, 0 \dots 0, 0)$$

$$g(2) = (0, 1, 0 \dots 0, 0)$$

$$\vdots$$

$$g(q-1) = (0, 0, 0 \dots 0, 1)$$

$$g(q) = (0, 0, 0 \dots 0, 0)$$

- ▶ **Abuse notation** and write $x = g(x) = (x_1, x_2, \dots, x_{q-1})$

$$P_{X,Y} \implies \alpha^{(i)} \implies \beta^{(i)} \implies \gamma^{(i)} \implies P_{X,Z,Y}^*$$

- ▶ We are given $P_{X,Y}$, where $|\mathcal{X}| = q$
- ▶ Need to produce $P_{X,Z,Y}^*$ by reducing to binary-input upgrading
- ▶ Denote $\mathcal{X}' = \{0, 1\}$
- ▶ Let $X = (X_1, X_2, \dots, X_{q-1})$ and Y be distributed according to $P_{X,Y}$
- ▶ **First step:** define, for $1 \leq i \leq q - 1$ the joint distribution

$$\alpha_{X_i,Y}^{(i)}(x', y) = P(X_i = x', Y = y | X_1^{i-1} = 0_1^{i-1})$$

- ▶ The joint distribution $\alpha_{X_i,Y}^{(i)}(x', y)$ has binary input, $x' \in \mathcal{X}'$
- ▶ We may apply our binary-input upgrading procedure

$$P_{X,Y} \implies \alpha^{(i)} \implies \beta^{(i)} \implies \gamma^{(i)} \implies P_{X,Z,Y}^*$$

- ▶ Recall our binary-input joint distribution: for $1 \leq i \leq q - 1$,

$$\alpha_{X_i,Y}^{(i)}(x', y) = P(X_i = x', Y = y | X_1^{i-1} = 0_1^{i-1})$$

- ▶ Define

$$\Lambda = \left\lfloor L^{1/(q-1)} \right\rfloor .$$

- ▶ **Second step:**

- ▶ Apply our **binary-input upgrading** procedure to $\alpha_{X_i,Y}^{(i)}(x', y)$, resulting in

$$\beta_{X_i,Z_i,Y}^{(i)}(x', z, y) ,$$

where

$$|Z_i| \leq \Lambda$$

- ▶ Difference in entropies is $O(\Lambda^{-2})$

$$P_{X,Y} \implies \alpha^{(i)} \implies \beta^{(i)} \implies \gamma^{(i)} \implies P_{X,Z,Y}^*$$

- ▶ Recall that we have produced $\beta_{X_i, Z_i, Y}^{(i)}(x', z, y)$, where $x' \in \mathcal{X}'$ is binary
- ▶ **Third step:** define the conditional distribution

$$\begin{aligned} \gamma_{X_i | Z_i, X_1^{i-1}}^{(i)}(x_i | z_i, x_1^{i-1}) \\ = \begin{cases} \beta_{X_i | Z_i}^{(i)}(x_i | z_i) & \text{if } x_1^{i-1} = 0_1^{i-1}, \\ 1 & \text{if } x_1^{i-1} \neq 0_1^{i-1} \text{ and } x_i = 0, \\ 0 & \text{otherwise.} \end{cases} \end{aligned}$$

- ▶ That is, if x_1^{i-1} is non-zero, force x_i to zero, in accordance with the one-hot representation
- ▶ Otherwise, if x_1^{i-1} is zero, use $\beta_{X_i | Z_i}^{(i)}$

$$P_{X,Y} \implies \alpha^{(i)} \implies \beta^{(i)} \implies \gamma^{(i)} \implies P_{X,Z,Y}^*$$

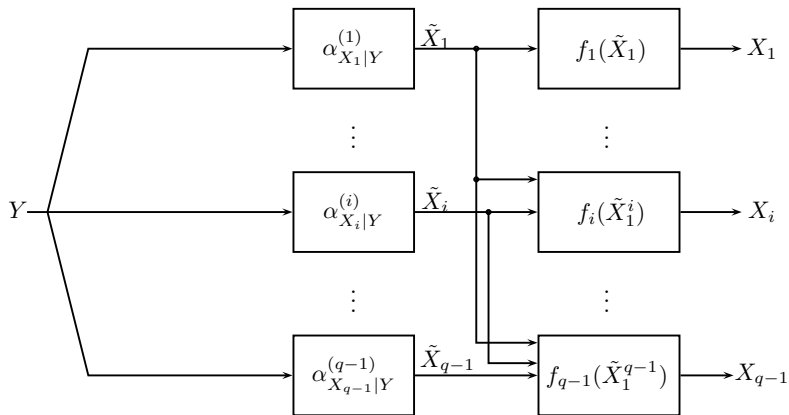
- ▶ Last step: define

$$P_{X,Z,Y}^*(x, z, y) = P_Y(y) \cdot \left(\prod_{i=1}^{q-1} \beta_{Z_i|Y}^{(i)}(z_i|y) \right) \cdot \left(\prod_{i=1}^{q-1} \gamma_{X_i|Z_i, X_1^{i-1}}^{(i)}(x_i|z_i, x_1^{i-1}) \right)$$

- ▶ A valid upgrade, with optimal power law:

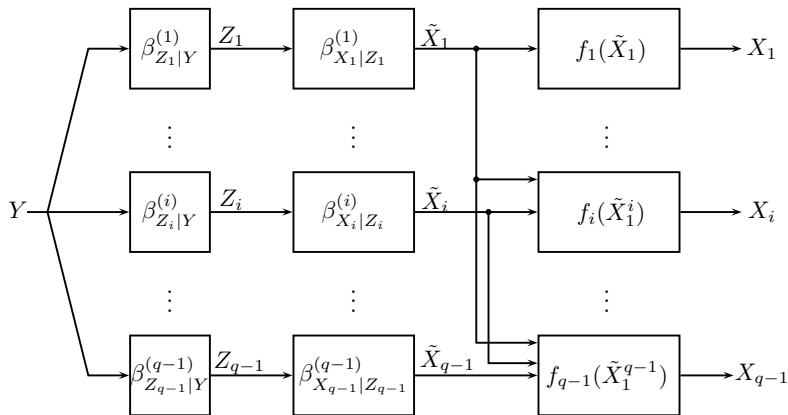
$$H(X|Y) - H(X|Z) = O(L^{-2/(|\mathcal{X}|-1)})$$

A graphical description of $P_{X,Y}$



where $f_i(\tilde{X}_1^i) \triangleq \tilde{x}_i \cdot \mathbb{1}_{\{\tilde{x}_1^{i-1} = \mathbf{0}_1^{i-1}\}}$

A graphical description of $P_{X,Z,Y}$



where $f_i(\tilde{X}_1^i) \triangleq \tilde{X}_i \cdot \mathbb{1}_{\{\tilde{X}_1^{i-1} = \mathbf{0}_1^{i-1}\}}$