

# **Universal Decoding for Asynchronous Slepian–Wolf Encoding**

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# Asynchronous Slepian–Wolf (ASW) Encoding

ASW encoding – **unknown** relative delay:  $X_i$  – correlated to  $Y_{i+\textcolor{red}{d}}$ .

Reasoning:

- ♣ No information exchange between the encoders.
- ♣ Different processing delays at the two encoders.
- ♣ Sampling clocks at the two locations are not synchronized.
- ♣ Relative delays between measurements due to different distances.

# Earlier Work

- ♠ Willems ('88): delay unknown to encoders, known to decoder.
- ♠ Rimoldi & Urbanke ('77); Sun *et al.* ('10): source splitting; decoder waits.
- ♠ Matsuta & Uyematsu ('20): worst-case approach:

The joint distribution,  $P_{XY}$ , is only known to be in  $\mathcal{S}$ .

Relative delay,  $d$ , is only known to be between  $-\Delta n$  and  $\Delta n$ ,  $\Delta \in [0, 1)$ .

Their main result:

$$R_x \geq \sup_{P_{XY} \in \mathcal{S}} [H(X|Y) + \Delta \cdot I(X;Y)]$$

$$R_y \geq \sup_{P_{XY} \in \mathcal{S}} [H(Y|X) + \Delta \cdot I(X;Y)]$$

$$R_x + R_y \geq \sup_{P_{XY} \in \mathcal{S}} [H(X, Y) + \Delta \cdot I(X;Y)]$$

## Earlier Work (Cont'd)

The worst case approach is **pessimistic**:

For example, if  $\mathcal{S}$  = the entire simplex of distributions over  $\mathcal{X} \times \mathcal{Y}$ ,

$$R_x \geq \log |\mathcal{X}|$$

$$R_y \geq \log |\mathcal{Y}|$$

$$R_x + R_y \geq \log |\mathcal{X}| + \log |\mathcal{Y}|$$

which is an **uninteresting triviality**.

Their analysis of the probability of error is also for the **worst** source in  $\mathcal{S}$ .

Q1: But what if  $\{P_{XY}\}$  is **not** the worst source in  $\mathcal{S}$ ?

Q2: Can one devise a **universal decoder** for unknown  $P_{XY}$  and  $d$ ?

# Earlier Work on Universal Decoding

## Universal channel decoding:

- ◊ Goppa ('75) - the MMI decoder achieves capacity.
- ◊ Csiszár & Körner ('81) – MMI decoder achieves  $E_r(R)$ .
- ◊ Csiszár ('82); Ziv ('85); Feder & Lapidot ('98); Feder & Merhav ('02).
- ◊ ...

## Universal S–W decoding:

- ◊ Csiszár & Körner ('81): minimum entropy decoder.
- ◊ Oohama & Han ('94); Draper ('04); Chen *et al.* ('08); Merhav ('16),...

# Main Contributions in This Work

- ♣ Proposing a universal decoder for unknown source and **delay**.
- ♣ Same random coding exponent as the MAP decoder.
- ♣ The lower bound is based on a lower bound to  $\Pr\{\cup_{i,j} \mathcal{A}_i \cap \mathcal{B}_j\}$ .
- ♣ Deriving a Lagrange–dual formula of the error exponent.
- ♣ Outlining a possible extension to sources with memory.

# Problem Formulation

- ♠  $\{(X_i, Y_{i+d})\}$  are i.i.d. pairs of finite-alphabet RVs  $\sim P_{XY}$ .
- ♠  $P_d(\mathbf{x}, \mathbf{y}) = \prod_{i=1}^d P_Y(y_i) \cdot \prod_{i=n-d+1}^n P_X(x_i) \cdot \prod_{i=1}^{n-d} P_{XY}(x_i, y_{i+d})$ .
- ♠ Encoder  $X$ :  $f(\mathbf{x}) \sim \text{unif}\{1, 2, \dots, 2^{nR_X}\}$ .
- ♠ Encoder  $Y$ :  $g(\mathbf{y}) \sim \text{unif}\{1, 2, \dots, 2^{nR_Y}\}$ .
- ♠ Neither  $P_{XY}$  nor  $d = \delta n$  are known.
- ♠ MAP decoder:  $(\hat{\mathbf{x}}, \hat{\mathbf{y}}) = \arg \max_{\mathbf{x}', \mathbf{y}'} P_d(\mathbf{x}', \mathbf{y}')$
- ♠ General metric decoder:  $(\hat{\mathbf{x}}, \hat{\mathbf{y}}) = \arg \max_{\mathbf{x}', \mathbf{y}'} q(\mathbf{x}', \mathbf{y}')$ .
- ♠ Error probability:  $P_E = \Pr\{(\hat{\mathbf{x}}, \hat{\mathbf{y}}) \neq (\mathbf{x}, \mathbf{y})\}$ .
- ♠ Error exponent:  $E(R_X, R_Y) = \lim_{n \rightarrow \infty} -\frac{\log P_E}{n}$ .

# The Proposed Universal Decoding Metric

For  $0 \leq k \leq n$ , define

$$u_k(\mathbf{x}, \mathbf{y}) = k\hat{H}(y_1^k) + (n - k)\hat{H}(x_1^{n-k}, y_{k+1}^n) + k\hat{H}(x_{n-k+1}^n),$$

$$v_k(\mathbf{x}, \mathbf{y}) = (n - k)\hat{H}(x_1^{n-k}|y_{k+1}^n) + k\hat{H}(x_{n-k+1}^n),$$

$$w_k(\mathbf{x}, \mathbf{y}) = k\hat{H}(y_1^k) + (n - k)\hat{H}(y_{k+1}^n|x_1^{n-k}),$$

$$q_k(\mathbf{x}, \mathbf{y}) = \max\{\textcolor{red}{u}_k(\mathbf{x}, \mathbf{y}) - n(R_x + R_y), \textcolor{blue}{v}_k(\mathbf{x}, \mathbf{y}) - nR_x, \textcolor{green}{w}_k(\mathbf{x}, \mathbf{y}) - nR_y\},$$

and finally, the universal decoding metric,  $q$ , is defined as

$$q(\mathbf{x}, \mathbf{y}) = \min_{0 \leq k \leq n} q_k(\mathbf{x}, \mathbf{y}).$$

# Main Theorem

$$E_{\text{univdec}}(R_x, R_y) = E_{\text{MAP}}(R_x, R_y) = \min\{E_{x|y}(R_x), E_{y|x}(R_y), E_{xy}(R_x, R_y)\},$$

where

$$E_{x|y}(R_x) = \max_{0 \leq \rho \leq 1} \rho \cdot [R_x - \delta H_{1/(1+\rho)}(X) - (1 - \delta) H_{1/(1+\rho)}(X|Y)]$$

$$E_{y|x}(R_y) = \max_{0 \leq \rho \leq 1} \rho \cdot [R_y - \delta H_{1/(1+\rho)}(Y) - (1 - \delta) H_{1/(1+\rho)}(Y|X)]$$

$$E_{xy}(R_x, R_y) = \max_{0 \leq \rho \leq 1} \rho \cdot [R_x + R_y -$$

$$\delta H_{1/(1+\rho)}(X) - \delta H_{1/(1+\rho)}(Y) - (1 - \delta) H_{1/(1+\rho)}(X, Y)].$$

# Discussion

- ♥ Three types of errors:  $X$  only,  $Y$  only, and both  $X$  and  $Y$ .
- ♥ The decoding metric has three components correspondingly.
- ♥ Uncertainty in  $d$  - treated differently than that of  $P_{XY}$ .
- ♥ Unconditional Rényi entropies correspond to independent segments.
- ♥ Achievable rate region for error exponent  $E$ :

$$\mathcal{R}(E) = \{(R_x, R_y) : R_x \geq \mathbf{R}_x(E), R_y \geq \mathbf{R}_y(E), R_x + R_y \geq \mathbf{R}_{xy}(E)\},$$

$$\mathbf{R}_x(E) = \inf_{s \geq 1} [sE + \delta H_{s/(1+s)}(X) + (1 - \delta) H_{s/(1+s)}(X|Y)]$$

and similar expressions for  $\mathbf{R}_y(E)$  and  $\mathbf{R}_{xy}(E)$ .

- ♥ Proof: upper bound – MoT. Lower bound – de Caen's inequality.
- ♥ Extension to Markov: replace empirical entropies by LZ metrics.

# Conclusion

- ♠ Replacing the worst-case approach by an “adaptive” approach.
- ♠ Achieves not only the rate region, but also the error exponent.
- ♠ Non-trivial universality in  $d$ .
- ♠ Universal metric with three components, one for every type of error.
- ♠ Characterization of rate region for a given error exponent,  $E$ .
- ♠ Lower bound on  $P[\cup_{i,j} \mathcal{A}_i \cap \mathcal{B}_j]$ : could be useful elsewhere too.