

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+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 Δn , $\Delta \in [0, 1)$.

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 & Lapidoth ('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

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

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

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

$$q_k(\mathbf{x}, \mathbf{y}) = \max\{u_k(\mathbf{x}, \mathbf{y}) - n(R_x + R_y), v_k(\mathbf{x}, \mathbf{y}) - nR_x, w_k(\mathbf{x}, \mathbf{y}) - nR_y\},$$

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.