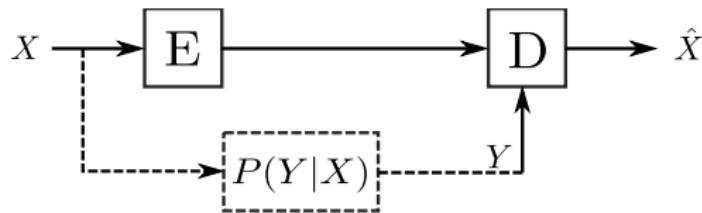


# Source Coding with Side Information at the Decoder: Models with Uncertainty, Performance Bounds, and Practical Coding Schemes

Elsa DUPRAZ<sup>†</sup> Aline ROUMY<sup>+</sup> and Michel KIEFFER<sup>†</sup>

<sup>†</sup> LSS - CNRS - SUPELEC - Univ Paris-Sud  
+ INRIA Rennes

# Introduction



- If  $P(X, Y)$  perfectly known
  - Performance :  $R_{X|Y}^{\text{SW}} = H(X|Y)$  bits/symbol
  - Practical Solution : LDPC codes
- If  $P(X, Y)$  not perfectly known
  - Performance : ?
  - Practical Solution : ?

# Outline

1 Introduction

2 Signal Models

3 Time Invariant Parameters

4 Time Varying Parameters

5 Conclusion

## 1 Introduction

## 2 Signal Models

## 3 Time Invariant Parameters

## 4 Time Varying Parameters

## 5 Conclusion

# Time Invariant Parameters

Unknown parameter  $\theta \in \mathcal{P}_\theta$  fixed for a sequence  $\{X_n, Y_n\}_{n=1}^{+\infty}$ ,  
varying from sequence to sequence

## Models

- P-Source :  $(X, Y) \sim P(X, Y|\theta)$ , prior  $\Theta \sim P_\Theta(\theta)$
- WP-Source :  $(X, Y) \sim P_\theta(X, Y)$

Stationary, non-ergodic models

Example :  $X$  BSS,  $Y|X$  BSC( $\theta$ )

# Time Varying Parameters

Unknown sequence of parameters  $\{\pi_n\}_{n=1}^{+\infty}$ ,  $\pi_n \in \mathcal{P}_\pi$

## Models

- M-Source :  $(X_n, Y_n) \sim P(X_n, Y_n | \pi_n)$ , prior  $\Pi \sim P_\Pi(\pi)$
- WPM-Source :  $(X_n, Y_n) \sim P_{\pi_n}(X_n, Y_n)$

M-Source stationary and ergodic

WPM-Source non-stationary and non-ergodic

Example :  $X_n$  BSS,  $Y_n | X_n$  BSC( $\pi_n$ )

## 1 Introduction

## 2 Signal Models

## 3 Time Invariant Parameters

## 4 Time Varying Parameters

## 5 Conclusion

# Performance

- **P-Source:** Information Spectrum approach [Han03]

$$R_{X|Y}^P = P_{\Theta}\text{-ess. sup } H(X|Y, \Theta = \theta)$$

- **WP-Source:** Universal Slepian-Wolf [Csi82]

$$R_{X|Y}^{WP} = \sup_{\theta \in \mathcal{P}_{\theta}} H_{\theta}(X|Y)$$

## Practical coding scheme (P-Source)

$X : \text{BSS}$ ,  $Y|X : \text{BSC}(\theta)$ , sequence  $\mathbf{X}_1^N$  to encode as  $(\mathbf{X}_1^{u_N}, \mathbf{Z}_1^M)$

- First step :

Learning sequence  $\mathbf{X}_1^{u_N}$

- Second step :

LDPC coding of  $\mathbf{X}_{u_N+1}^N$  as  $\mathbf{Z}_1^M = H^T \mathbf{X}_{u_N+1}^N$

where  $H$  : degree distributions  $(\lambda(x), \rho(x))$

### Question

How to choose the length  $u_N$  of the learning sequence?



# Practical coding scheme (P-Source)

The LDPC decoder requires  $\theta = P(X = 1 | Y = 0)$

- **Decoding** : LDPC decoder initialized with

$$\theta_r := P(X_n = 1 | Y_n = 0, \mathbf{x}_1^{u_N}, \mathbf{y}_1^{u_N}) = \int_{\theta \in \mathcal{P}_\theta} P(X_n | Y_n, \theta) P_{\Theta | \mathbf{x}_1^{u_N}, \mathbf{y}_1^{u_N}}(\theta) d\theta$$

mismatch on the parameter ( $\theta_r \neq \theta$ )

- **Performance** : Density Evolution initialized with

$$p_X(x) = \theta \delta \left( x + \log \frac{1 - \theta_r}{\theta_r} \right) + (1 - \theta) \delta \left( x - \log \frac{1 - \theta_r}{\theta_r} \right)$$

# Practical coding scheme (P-Source)

$\gamma$  : outage measure

- ① Outage set  $B_{\theta|\theta_r}^\gamma$  s.t.  $\int_{\theta \in B_{\theta|\theta_r}^\gamma} P_{\Theta|\theta_r}(\theta) d\theta > 1 - \gamma$
- ② Error Probability :  $P_e(\lambda, \rho, \theta, \theta_r)$
- ③ Set of admissible parameters  $(\lambda, \rho)$  :

$$\Gamma(\gamma, \epsilon_t) = \{(\lambda, \rho) \text{ s.t. } \forall \theta_r \in \text{Conv}(\mathcal{P}_\theta), \exists B_{\theta|\theta_r}^\gamma : \forall \theta \in B_{\theta|\theta_r}^\gamma, P_e(\lambda, \rho, \theta, \theta_r) < \epsilon_t\}$$

Total rate

$$R^N(\gamma, \epsilon_t) = \frac{u_N}{N} H(X) + \frac{N - u_N}{N} \inf_{(\lambda, \rho) \in \Gamma(\gamma, \epsilon_t)} R(\lambda, \rho)$$

## 1 Introduction

## 2 Signal Models

## 3 Time Invariant Parameters

## 4 Time Varying Parameters

## 5 Conclusion

# Performance

- M-Source: Classical Slepian-Wolf [SW73]

$$R_{X|Y}^M = H(X|Y)$$

- WPM-Source: Arbitrarily Varying Sources [Ahl79]

$$R_{X|Y}^{WPM} = \sup_{P(X,Y) \in \text{Conv}(\{P_\pi(X,Y)\}_{\pi \in \mathcal{P}_\pi})} H(X|Y)$$

# Practical coding scheme

- M-Source : Standard soft LDPC decoder :

$$P(X, Y) = \sum_{\pi \in \mathcal{P}_\pi} P(\pi) P(X, Y | \pi).$$

- WPM-Source : Standard hard LDPC decoder  
Loss in performance

# Experiments

**Table:** Theoretical (Th.) and practical (Prac.) rate bounds in bit/symbol when  $X$  is a BSS and  $P(Y|X)$  a BSC.

Source	Conditions	Th. Rate	Prac. Rate
P-Source	$\mathcal{P}_\theta = [0.1, 0.21]$	0.74	0.75
WP-Source	$\mathcal{P}_\theta = [0.1, 0.21]$	0.74	0.75
M-Source	$\mathcal{P}_\pi = \{0.1, 0.21\}$ $p = 0.143$	0.59	0.6
WPM-Source	$\mathcal{P}_\pi = \{0.1, 0.143\}$	0.59	0.75
No SI	$P(X = 1) = 0.5$	1	1

P,WP-Sources :  $(0.093x^3 + 0.720x^4 + 0.187x^5, x^5)$

M-Source :  $(0.099x^3 + 0.712x^4 + 0.174x^5 + 0.015x^6, x^5)$

WPM-Source : (3,4)

# Conclusion and Perspectives

- Practical coding scheme for each model
- Analysis to extend to the non-binary case
- How to avoid the learning sequence?
- How to build a coding scheme for the WPM-Source that performs close to the entropy?

# Bibliography

- [Ahl79] R. Ahlswede.  
Coloring hypergraphs: A new approach to multi-user source coding-1.  
*Journal of Combinatorics*, 4(1):76–115, 1979.
- [Csi82] I. Csiszar.  
Linear codes for sources and source networks: Error exponents, universal coding.  
*IEEE Trans. on Inf. Th.*, 28(4):585–592, 1982.
- [Han03] T.S. Han.  
*Information-spectrum methods in information theory*.  
Springer, 2003.
- [SW73] D. Slepian and J. Wolf.  
Noiseless coding of correlated information sources.  
*IEEE Trans. on Inf. Th.*, 19(4):471–480, July 1973.