

---

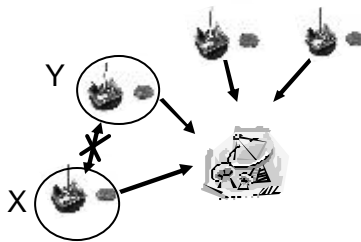
# **Distributed Source Coding: A New Paradigm for Wireless Video ?**

Christine Guillemot, IRISA/INRIA, Campus universitaire de Beaulieu, 35042 Rennes Cédex, FRANCE  
Christine.Guillemot@irisa.fr



# The distributed sensor network problem

- How to compress multiple correlated sensor outputs that do not communicate with each other?
  - capturing the redundancy in dense sensor networks

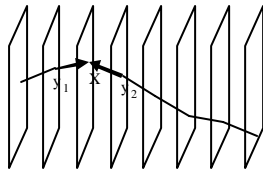


- Let us consider two sensors X, Y capturing correlated data:
  - **Can we exploit correlation between X & Y without communicating between the two sensor nodes?**
    - o Communication between nodes consume energy and bandwidth

# Video Compression so far ...

## Applications

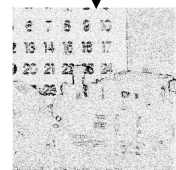
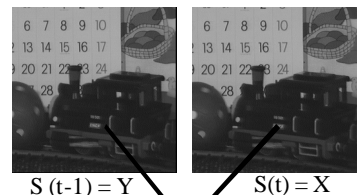
- o Storage (CD, DVD)
- o Broadcasting
- o Streaming video-on-demand
- o ...



- Motion-based search of best predictor by coder
- Computes a prediction error
- Motion fields and prediction error transmitted
- Given the motion fields, the decoder can find the predictors

The answer today:

Motion-based predictive coding  
(MPEG-x, H.26x)



$$E_p = X - f(\hat{Y}_1, \dots, \hat{Y}_M)$$

**High compression efficiency, High encoding complexity**  
**High sensitivity to transmission noise**

# The wireless video problem ...

## analogy with the sensor network problem

### Wireless scenario

- o Wireless digital cameras
- o Mobile phones, PDA's
- o Low-power video sensors
- o Wireless video teleconferencing



### Challenges:

- Narrow bandwidth
  - high compression efficiency
- Limited handheld battery power
  - low end-device (encoder) complexity
- Lossy & erroneous medium
  - Robustness to transmission impairments

### Re-thinking the classical motion-based predictive coding paradigm !

- Adjacent frames as correlated 2D camera sensor data
- X, Y modelled as sequences of correlated Random Variables.

- Can the encoder compress X without knowing the realization of Y, but only the pmf(X|Y), with performances matching the classical predictive paradigm?
- Separate coding => decreased encoder complexity, in-built resilience

# Outline

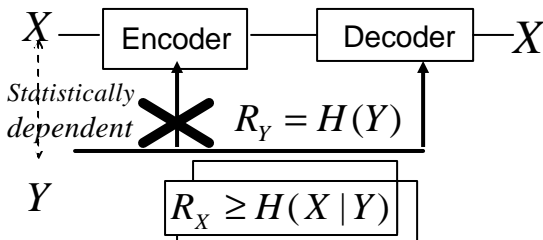
---

- Asymptotic answers from information theory
- Constructive solutions based on the analogy with the channel coding problem
- A range of applications
- Application to wireless video compression

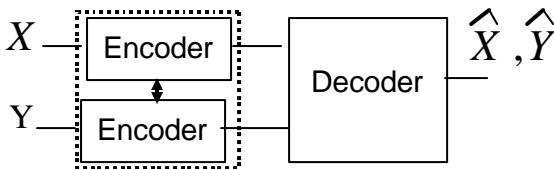
# Asymptotic answers

## Lossless Case

### Coding with side information



### Slepian-Wolf theorem (1973)



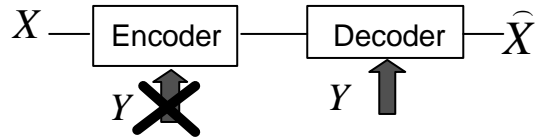
$$R_X + R_Y = H(X, Y)$$

with vanishing error probability if sequences Sufficiently long.



## Lossy Case

### Wyner-Ziv Theorem (1976)



$$R \geq R_X(D_X)$$

$$R \geq R_{X|Y}(D_X)$$

$$R_X \geq R_{X|Y}^{wz}(D_X) \geq R_{X|Y}(D_X)$$

(Rate loss)

Equality under the Gaussianity assumption and with MSE as distortion measure

**Don't forget: X and Y will be adjacent frames in the sequence!!**



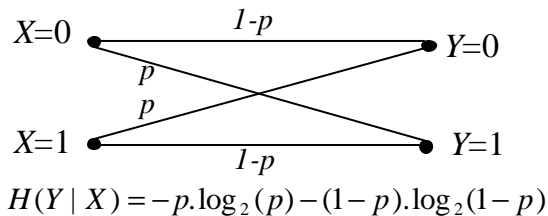
---

Constructive solutions  
based on the analogy with the channel  
coding problem

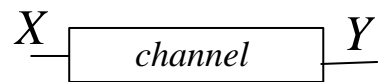
# Analogy with the channel coding problem

- The correlation between  $X$  and  $Y$  can be modeled as a *virtual channel*

Examples: **Binary Symmetric virtual Channel**

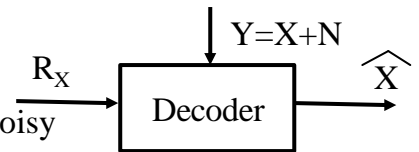


**AWGN virtual Channel**



$$Y = X + N ; P(Y/X)$$

- Side Information  $Y$  seen as a noisy observation of  $X$
- Minimum number of bits  $R_X$  to estimate  $X$  from the noisy observation  $Y$



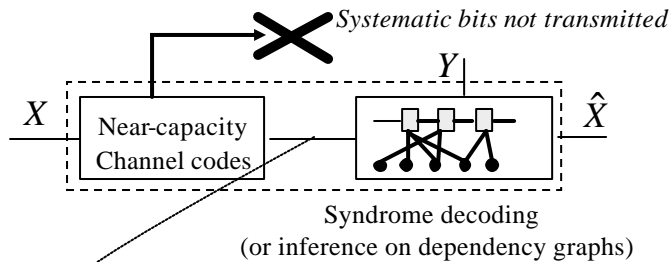
- Near-capacity channel codes to approach the S-W limit



# Slepian-Wolf Coder/Decoder Design

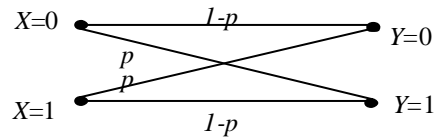
- Code adapted to the correlation virtual channel

- How to compress?: extract from X the minimum information so that the decoder can estimate X given Y

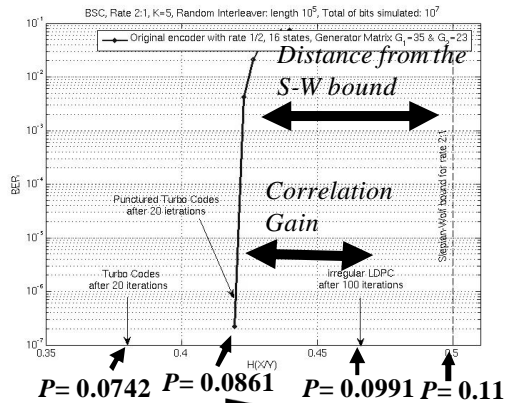


Only the parity bits are transmitted  
 $n \Rightarrow n-k$  bits

Compressing X at  $R_x=0.5$   
 Measuring  $H(X/Y)$   
 for vanishing error probability  
 $H(X/Y) = 0.4233$  for  $p=0.08$

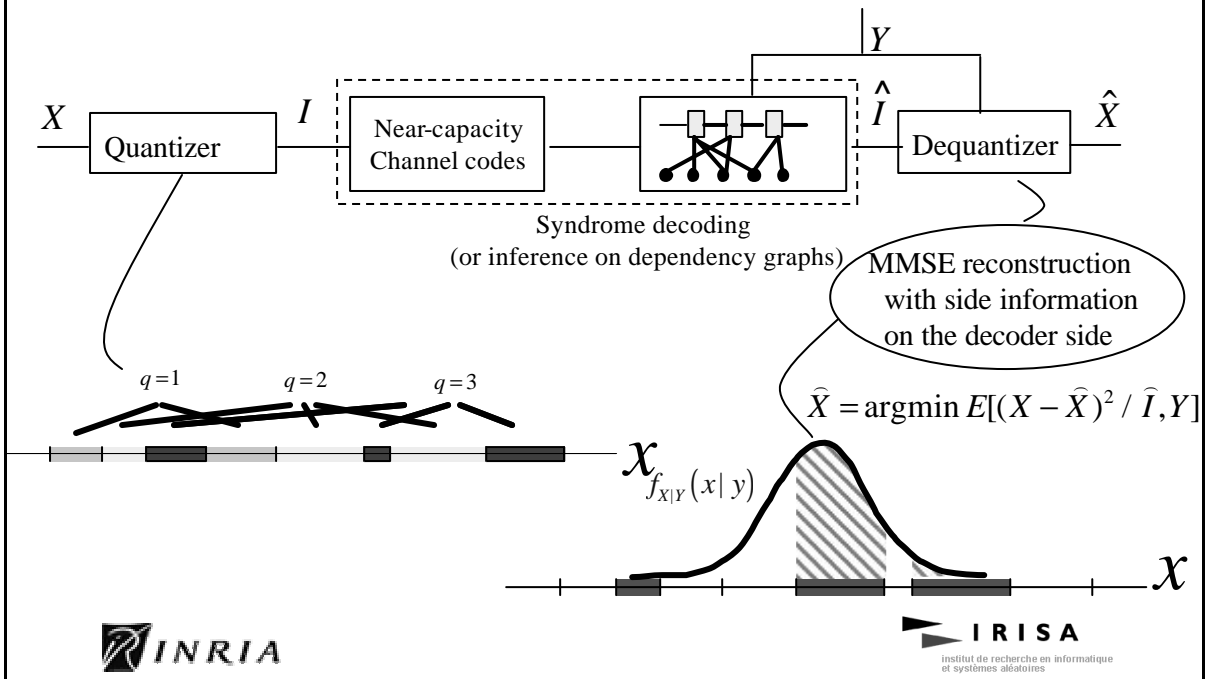


For  $p=0.11$ , S-W bound:  $H(X/Y)=0.5$



# Wyner-Ziv Coder/Decoder Design

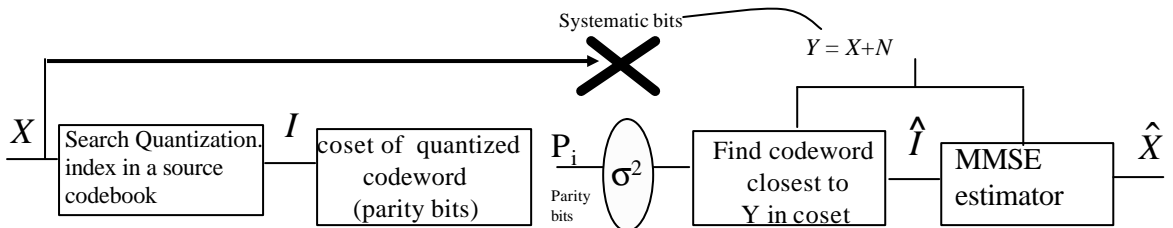
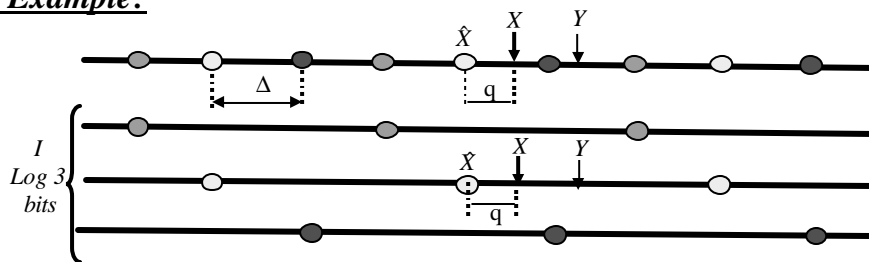
- Design a source codebook achieving the granular gain



# Optimum Design of W-Z Coder/Decoder: A Set Partitioning Problem

- Partitioning of the source codebook into a *good* channel code

## Toy Example:



$$\hat{X} = \operatorname{argmin} E[(X - \hat{X})^2 / \hat{I}, Y]$$

# A Few References ...

---

## ❑ With coset and multi-level coset codes

- Ramchandran & Pradhan, DISCUS 99
- Majumdar, Chou & Ramchandran 03, ...

## ❑ With Turbo Codes (two or multiple binary and/or Gaussian sources)

- Garcia-Frias & Zhao 01
- Bajcsy & Mitran 01
- Aaron & Girod 02
- Liveris & Xiong 02
- Lajnef, Guillemot & Siohan 04, ...

## ❑ With Low Density Parity Check Codes

- Liveris, Xiong & Georghiades 02,
- Ramchandran & Pradhan 02,
- Garcia-Frias & Zhong 03, ...

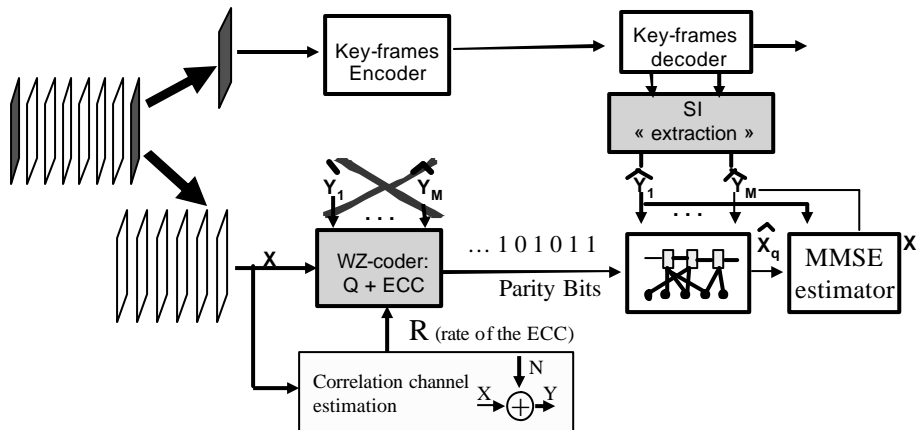
# Framework with a Range of Applications

---

- Distributed (dense) sensor networks
- Compression in embedded environments, e.g., hyperspectral compression for satellite imagery [Cheung, Wang, Ortega 05]
- Wireless video**
- M-channel Multiple description coding [Puri et al. 04]
  - drift correction with low latency in video communication
- Multimedia security: data hiding, watermarking, steganography
  - duality between source coding and channel coding with side information
- Compression of encrypted data [Johnson et al. 04]

# Distributed Coding for Wireless Video

- ❑ Low encoding complexity, No drift
  - Each frame is compressed independently (assuming only pmf  $(X, Y)$ )
    - Motion-free, prediction-free transmission => no drift, error-resilience
  - Side-information  $Y$  is exploited at the decoder ONLY

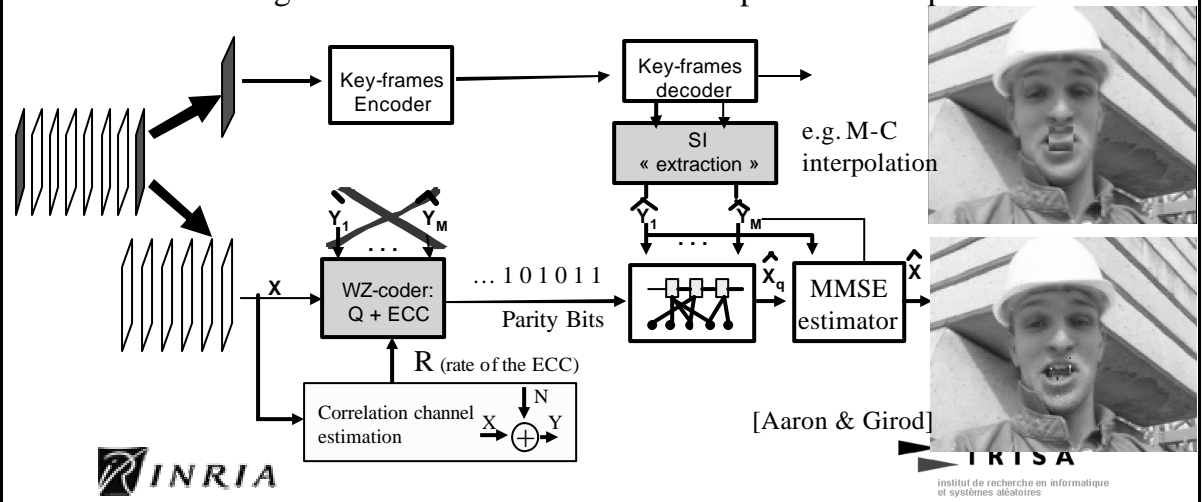


# Distributed Coding for Wireless Video: .... A number of open issues

- ❑ No MSE performance loss over case when  $Y$  is available at both encoder and decoder **if the innovation of  $X$  w.r.t.  $Y$  is Gaussian**

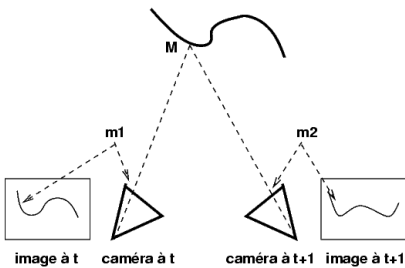
**ISSUE 1: Extraction of side information approaching at best this Gaussianity assumption AND with maximum correlation**

One solution: A good motion model + Motion-compensated Interpolation

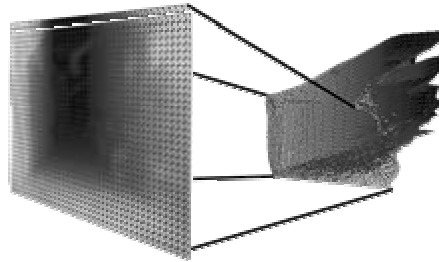


# Side information extraction

- At the crossroad of information theory, signal processing and computer vision
  - Capturing the correlation via scene geometrical constraints
    - 3D reconstruction with epipolar constraints and 2D projection
  - Some sort of motion estimation at the decoder
    - In theory *joint typicality*, in practice extra information as a CRC, hash function, ...



$$\min_{M_i, R_j, t_j} \sum_j \sum_i \|m_i^j - A.(R_j | t_j).M_i\|$$



Mesh-based dense motion field

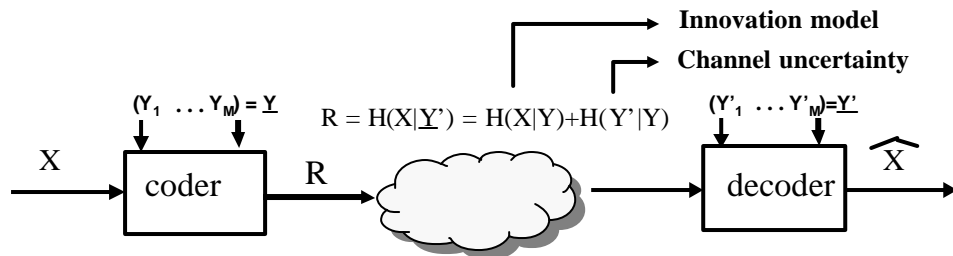
Depth Map

3D model



# Distributed Coding for Wireless Video

- ❑ Codes depend on the *virtual correlation channel* between X and Y
  - **ISSUE 2: Correlation estimation/tracking with minimum information exchange**



- Feature points extraction and tracking for camera position estimation
- Improving the quality of the 2D projection

Feature points tracking  
[Maitre, Morin, Guillemot]



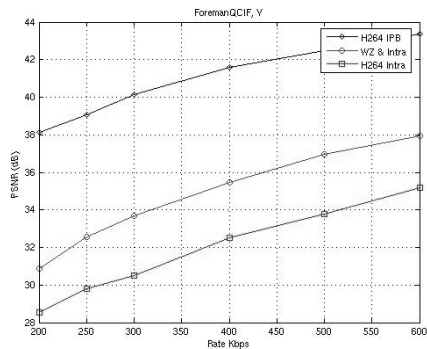
INRIA - Institut National de Recherche en Informatique et systèmes algorithmiques

# Performance illustrations

## H.264 standard



## H.264 Intra Mode (QP fixed)



## W-Z coder/decoder



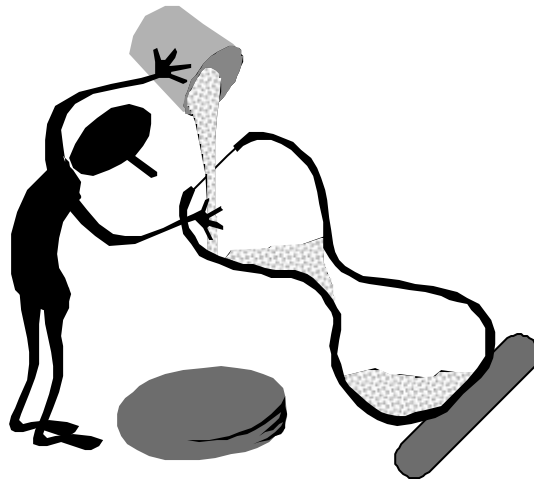
# Concluding remarks

---

- Re-thinking the compression paradigm
  - Signal compression based on error correcting codes
  - A joint source-channel coding problem
- Features for wireless video (mobile video cameras, uplink wireless video)
  - Complexity balancing between coder and decoder
  - In-built resilience to channel impairments (e.g., « motion-free » video coding and transmission).
- A number of open issues for real systems

# Thanks for listening!

---



Acknowledgements: V. Chappelier, H. Jégou, K. Lajnef, L. Morin (IRISA)  
M. Maître (Univ. Illinois at Urbana Champaign)  
P. Siohan (France Télécom)

