

## Lecture 20

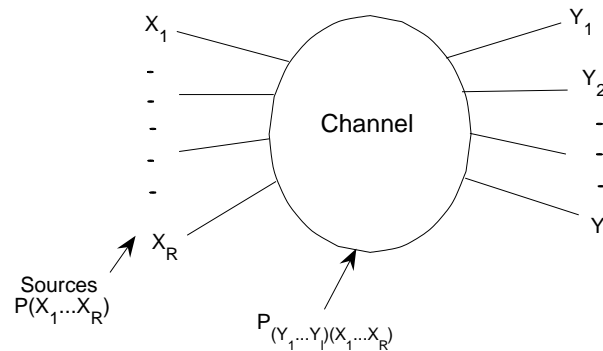
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## 1 Overview

In this lecture, we will continue with the theme of network information theory.

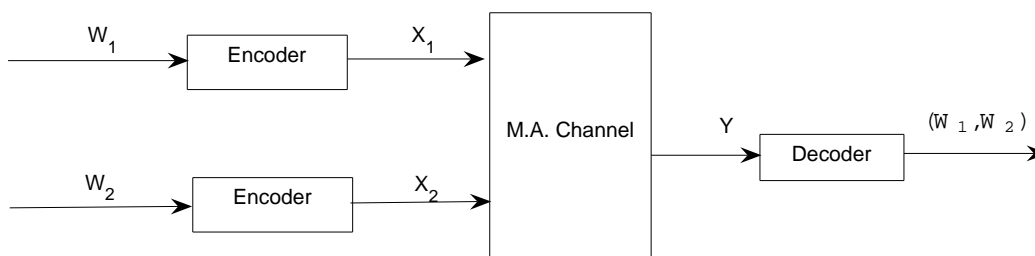
- Correlated-Sources Coding
- Side Information (an aside)
- Broadcast Channel (We ran out of time and this will be covered in next lecture)



Channel is characterized by transition probabilities.

$R_{ij}$  = requested rate from  $X_i \rightarrow Y_j$

### MULTIPLE ACCESS



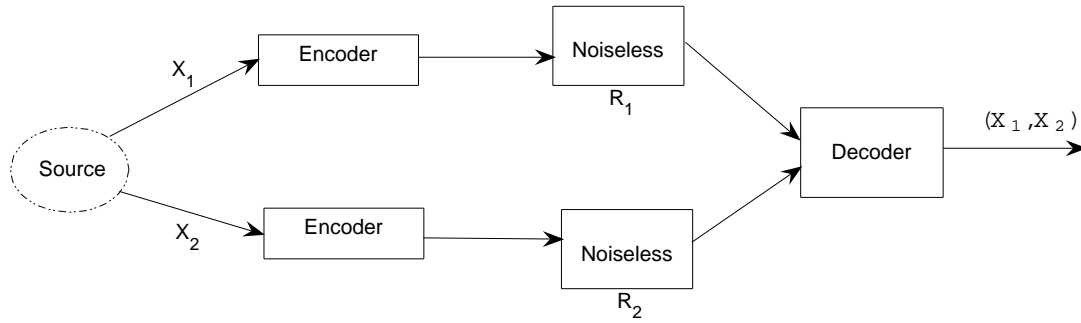
## 2 Correlated Source Coding

Where the arrows indicate how “hard” either case is (increasing in arrow direction).

Back to the Correlated Source Coding problem:

Definition:  $(R_1, R_2)$  achievable if  $\exists f_1, f_2$

$$\begin{cases} f_1 : \Omega_{X_1}^n \rightarrow \{1 \dots 2^{nR_1}\} \\ f_2 : \Omega_{X_2}^n \rightarrow \{1 \dots 2^{nR_2}\} \end{cases}$$



	Channel	Source
M.A. Channel	Noisy ↑	Uncorrelated ↓
Correlated Source	Noiseless	Correlated

$$g : \{1 \dots 2^{nR_1}\} \times \{1 \dots 2^{nR_2}\} \rightarrow \Omega_{X_1}^n \times \Omega_{X_2}^n$$

$$(X_1, X_2) \rightarrow (f_1(x), f_2(x)) \xrightarrow{g} (\hat{X}_1, \hat{X}_2)$$

such that

$$P_{err}^n = Pr[(X_1, X_2) \neq (\hat{X}_1, \hat{X}_2)] \xrightarrow{n} 0$$

$X_1$  is a source of entropy  $H(X_1) = H_1$ .

$X_2$  is a source of entropy  $H(X_2) = H_2$

$$I(X_1; X_2) = I$$

### 3 Slepian-Wolf Theorem

$(R_1, R_2)$  achievable iff

$$R_1 \geq H(X_1|X_2) = H_1 - I$$

$$R_2 \geq H(X_2|X_1) = H_2 - I$$

$$R_1 + R_2 \geq H(X_1, X_2) = H_1 + H_2 - I$$

#### ENCODING

Pick  $f_1$  at random

Pick  $f_2$  at random

Decoding $_{(f_1, f_2)}[Y_1, Y_2]$ : if  $\exists$  unique  $(\hat{X}_1, \hat{X}_2)$  such that :

①  $Y_1 = f_1(\hat{X}_1), Y_2 = f_2(\hat{X}_2)$ .

AND

② If  $(\hat{X}_1, \hat{X}_2)$  are jointly typical then output  $(\hat{X}_1, \hat{X}_2)$  else ERROR.

ANALYSIS:

Error Type 1 ( $X_1, X_2$ ) is not jointly typical:  $Pr \rightarrow 0$  (LLN).

Error Type 2:

Ⓐ  $\exists \hat{X}_1 \neq X_1, \hat{X}_2 \neq X_2$  such that  $(\hat{X}_1, \hat{X}_2)$  satisfy ① and ②.

To bound  $Pr[\text{Ⓐ}]$ :

Fix  $\hat{X}_1, \hat{X}_2, X_1, X_2$  with  $\hat{X}_1 \neq X_1$  and  $\hat{X}_2 \neq X_2$

$$Pr_{f_1, f_2}[f_1(\hat{X}_1) = f_1(X_1) \text{ and } f_2(\hat{X}_2) = f_2(X_2)] = 2^{-n(R_1 + R_2)}$$

Union bound over  $(\hat{X}_1, \hat{X}_2)$  jointly typical. Number of jointly typical  $(\hat{X}_1, \hat{X}_2) \leq 2^{n(H(X_1, X_2) + \epsilon)}$ .

If  $R_1 + R_2 > H(X_1, X_2)$  then  $Pr[\text{Ⓐ}] \rightarrow 0$ .

Ⓑ  $\exists \hat{X}_1 \neq X_1$  such that  $(\hat{X}_1, X_2)$  satisfy ① and ②.

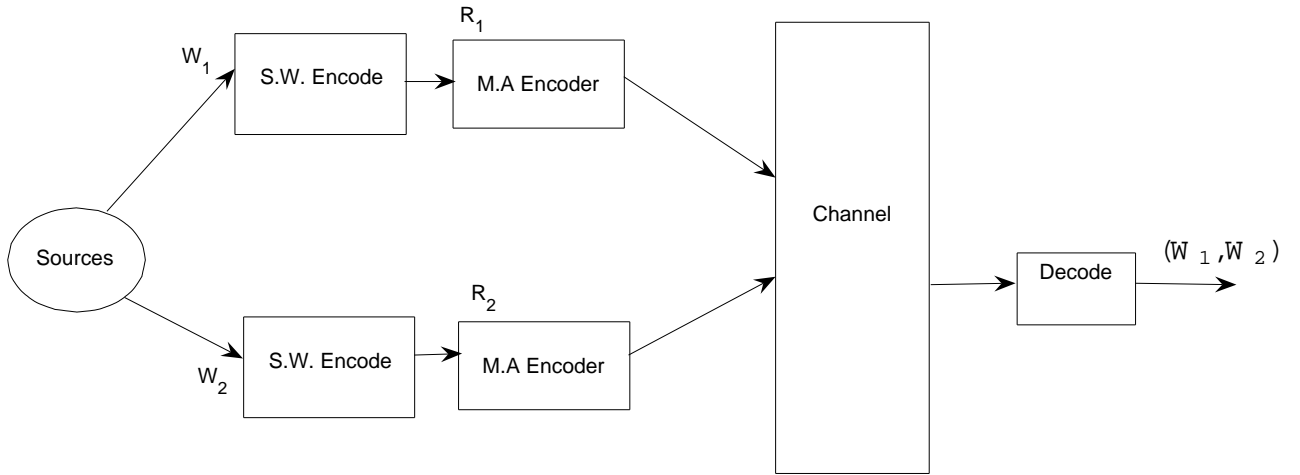
Fix  $\hat{X}_1, X_1$  with  $\hat{X}_1 \neq X_1$

$$Pr_{f_1, f_2}[f_1(\hat{X}_1) = f_1(X_1)] = 2^{-n(R_1)}$$

Union bound over  $(\hat{X}_1, X_2)$  jointly typical. Number of  $\hat{X}_1$  s.t.  $(\hat{X}_1, X_2)$  jointly typical =  $2^{nH(X_1|X_2)}$ .

$Pr[X_2] \leq 2^{-n(H(X_2) - \epsilon)}$ ,  $Pr[\hat{X}_1, X_2] \geq 2^{-n(H(X_1, X_2) + \epsilon)}$ . If  $R_1 > H(X_1|X_2)$  then  $Pr[\text{Ⓑ}] \rightarrow 0$ .

Ⓒ  $\exists \hat{X}_2 \neq X_2$  such that  $(X_1, \hat{X}_2)$  satisfy ① and ②. Similarly to Ⓑ: If  $R_2 > H(X_2|X_1)$  then  $Pr[\text{Ⓒ}] \rightarrow 0$ .



if  $\exists R_1, R_2$

$$R_1 \geq H(W_1|W_2)$$

$$R_2 \geq H(W_2|W_1)$$

$$R_1 + R_2 \geq H(W_1, W_2)$$

and  $(R_1, R_2)$  are achievable for MA channel  $(R_1, R_2) \in \text{convex hull}(\tilde{R}_1, \tilde{R}_2)$  s.t.  $P_{Y|(X_1, X_2)}$

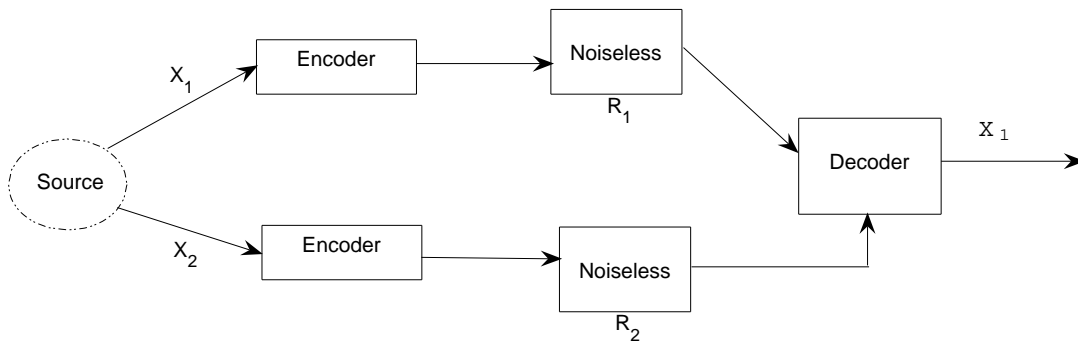
$$\tilde{R}_1 \leq I(X_1; Y|X_2)$$

$$\tilde{R}_2 \leq I(X_2; Y|X_1)$$

$$\tilde{R}_1 + \tilde{R}_2 \leq I(Y; (X_1, X_2))$$

the transmission is feasible.

## 4 Side Information



$X_2 \longrightarrow$  Decoder  
Suffices if:

$$\begin{aligned} R_1 &\geq H(X_1|X_2) \\ R_2 &\geq H(X_2|X_1) \\ R_1 + R_2 &\geq H(X_1, X_2) \end{aligned}$$

For example:  $X_1 = Z_1Z_2$  and  $X_2 = Z_2Z_3$ .  
Then:

$$\begin{aligned} R_1 &\geq H(X_1|X_2) = H(Z_1) \\ R_2 &\geq H(X_2|X_1) = H(Z_3) \\ R_1 + R_2 &\geq H(X_1, X_2) = H(Z_1) + H(Z_2) + H(Z_3) \end{aligned}$$

Instead, since we don't care about  $X_2$ :

$$\begin{aligned} R_1 &\geq H(Z_1) \\ R_2 &\geq 0 \\ R_1 + R_2 &\geq H(Z_1) + H(Z_2) = H(X_1) \end{aligned}$$

$(R_1, R_2)$  suffices if  $\exists \hat{X}_2$  s.t.  $X_1 \longrightarrow X_2 \longrightarrow \hat{X}_2$   
s.t.

$$\begin{aligned} R_1 &\geq H(X_1|\hat{X}_2) \\ R_2 &\geq H(\hat{X}_2|X_1) \\ R_1 + R_2 &\geq H(X_1, \hat{X}_2) \end{aligned}$$

What we want, or what would be nice to have is:

$$\begin{aligned} R_2 &\geq H(\hat{X}_2|X_1) = 0 \\ R_1 + R_2 &\geq H(X_1, \hat{X}_2) = H(X_1) \end{aligned}$$

OR

$$\begin{aligned} R_1 &\geq H(X_1|\hat{X}_2) \\ R_2 &\geq I(\hat{X}_2; X_1) \end{aligned}$$

THEOREM:

Side information problem is realizable for  $(X_1, X_2)$  if  $(R_1, R_2)$   
if  $\exists \hat{X}_2$  s.t.  $X_1 \longrightarrow X_2 \longrightarrow \hat{X}_2$

$$R_1 \geq H(X_1 | \hat{X}_2)$$

$$R_2 \geq I(X_1; \hat{X}_2)$$