

## Lecture 2: Some Information Theory

see: Classical Information Theory &  
Classical Error Correction

m: Lecture Notes on Quantum Information

D. Brups & G. Leuchs (see webpage)

### Def. Shannon entropy

Let  $S$  be a source emitting symbols  
 $x_1, \dots, x_n$  with probabilities  $p(x_i) = p_i$ ,  
with identical independent distribution (i.i.d).

Then the Shannon entropy of  $S$  is given by

$$H(S) = - \sum_{i=1}^n p(x_i) \log_2 p(x_i)$$

Independence  $p(x_i x_j) = p(x_i) p(x_j)$

$$H(S \times S) = - \sum_{i,j=1}^n p(x_i x_j) \log_2 p(x_i x_j)$$

$$= - \sum_{i,j} p_i p_j \log_2 (p_i p_j)$$

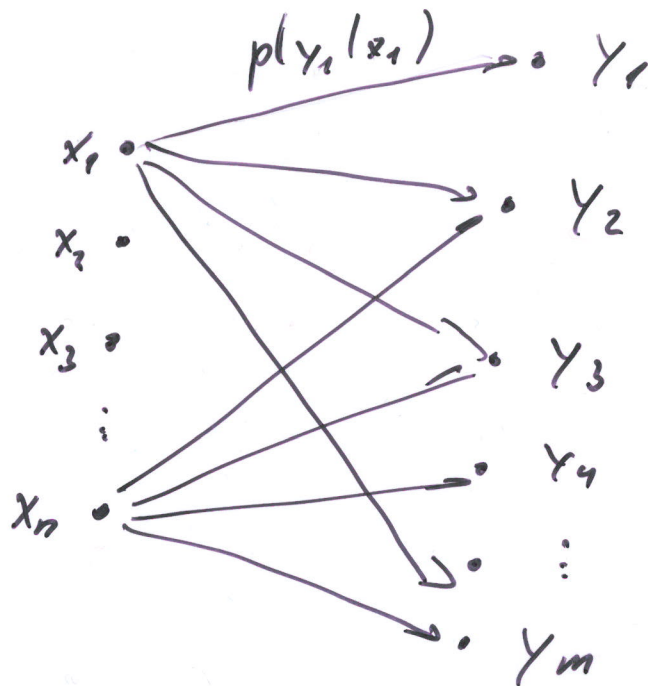
$$= - \sum_{i,j} p_i p_j \log_2 p_i - \sum_{i,j} p_i p_j \log_2 p_j$$

$$= - \sum_i p_i \log_2 p_i - \sum_j p_j \log_2 p_j$$

$$= 2 H(S)$$

recall:

channel with input alphabet  $X = \{x_1, \dots, x_n\}$   
and output alphabet  $Y = \{y_1, \dots, y_m\}$   
discrete memoryless channel described  
by transition probabilities  $p(y_j | x_i)$



edges are labeled by non-zero transition probs.

Def. joint entropy of the input  $X$  and output  $Y$

$$H(X, Y) = - \sum_{x \in X} \sum_{y \in Y} p(x, y) \log_2 p(x, y)$$

consider input and output as one symbol

Def.: conditional entropy

(3)

$$\begin{aligned} H(Y|X) &= \sum_{x \in \mathcal{X}} p(x) H(Y|X=x) \\ &= - \sum_{x \in \mathcal{X}} p(x) \sum_{y \in \mathcal{Y}} p(y|x) \log_2 p(y|x) \\ &= - \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} p(x,y) \log_2 p(y|x) \end{aligned}$$

$H(Y|X)$  is a measure for the information we additionally get when considering  $X$  and  $Y$  together, w/ only  $X$ .

Fact: chain rule

$$H(X, Y) = H(X) + H(Y|X)$$

Def.: mutual information

$$\begin{aligned} I(X; Y) &= H(X) + H(Y) - H(X, Y) \\ &= H(X) - H(X|Y) \\ &= H(Y) - H(Y|X) \end{aligned}$$

# Channel Capacity:

How much information can be sent reliably through a channel?

For a discrete memoryless channel, the capacity is:

$$C := \max_{p(x)} I(X; Y) = \max_{p(x)} (H(X) - H(X|Y))$$

where the maximization is over all input distributions.  
"encodings"

## Shannon's<sup>2nd</sup> second fundamental coding theorem

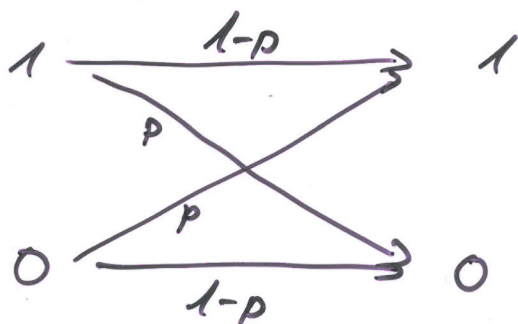
noisy coding theorem

Let  $S$  be a source with entropy  $H(S)$  and let a discrete memoryless channel have capacity  $C$ .

If  $H(S) < C$  then there exists an encoding scheme such that the output of the source can be transmitted over the channel with an arbitrary small probability of error. Not possible for  $H(S) > C$

# Example Capacity of the BSC

(5)



$$p(0,0) = p(1,1) = 1-p$$

$$p(0,1) = p(1,0) = p$$

$$\begin{aligned} I(X; Y) &= H(Y) - H(Y|X) \\ &= H(Y) - \sum p(x) H(Y|X=x) \\ &= H(Y) - \sum p(x) H(p) \\ &= H(Y) - H(p) \\ &\leq 1 - H(p) \end{aligned}$$

binary entropy function

$$H(p) := -p \log_2 p - (1-p) \log_2 (1-p)$$

