# Coding and Information Theory Chapter 4 Information Channels - B

Xuejun Liang Fall 2022

#### 4.2 The Binary Symmetric Channel

- Binary symmetric channel (BSC)
  - $A = B = Z_2 = \{0, 1\}.$
  - the channel matrix has the form

$$M = \begin{pmatrix} P_{00} & P_{01} \\ P_{10} & P_{11} \end{pmatrix} = \begin{pmatrix} P & \overline{P} \\ \overline{P} & P \end{pmatrix}$$

for some *P* where  $0 \le P \le 1$ 

The input probabilities have the form

$$p_0 = \Pr(a = 0) = p,$$
  
 $p_1 = \Pr(a = 1) = \overline{p},$ 

for some p such that  $0 \le p \le 1$ 

• The channel relationships for BSC = ?

$$q_0 = \Pr(b = 0) = pP + \overline{p}\overline{P},$$
  
 $q_1 = \Pr(b = 1) = p\overline{P} + \overline{p}P;$ 

$$\sum_{i=1}^r p_i P_{ij} = q_j$$

writing  $q_0 = q$  and  $q_1 = \overline{q}$  we then have

$$(q,\overline{q}) = (p,\overline{p}) \begin{pmatrix} P & \overline{P} \\ \overline{P} & P \end{pmatrix}$$

• Bayes' formula for BSC = ?

$$Q_{00} = \frac{pP}{pP + \overline{p}\overline{P}}, \quad Q_{10} = \frac{\overline{p}\overline{P}}{pP + \overline{p}\overline{P}}$$

$$Q_{01} = \frac{p\overline{P}}{p\overline{P} + \overline{p}P}, \quad Q_{11} = \frac{\overline{p}P}{p\overline{P} + \overline{p}P}$$

$$Q_{ij} = \frac{p_i P_{ij}}{\sum_{k=1}^r p_k P_{kj}}$$

#### • Example 4.4

- Let the input A be defined by putting p = 1/2
- Probabilities of the output symbols:  $q_0$  = ? And  $q_1$  = ?
- The backward probabilities:  $Q_{00}$ ,  $Q_{01}$ ,  $Q_{10}$ ,  $Q_{11}$ , = ?

$$q_0 = pP + \overline{p}\overline{P}$$

$$q_1 = p\overline{P} + \overline{p}P$$

$$Q_{00} = \frac{pP}{pP + \overline{p}\overline{P}} \quad Q_{11} = \frac{\overline{p}P}{p\overline{P} + \overline{p}P} \quad Q_{10} = \frac{\overline{p}\overline{P}}{pP + \overline{p}\overline{P}} \quad Q_{01} = \frac{p\overline{P}}{p\overline{P} + \overline{p}P}$$

$$Q_{ij} = \frac{p_i}{q_j} P_{ij}$$

- Suppose that P = 0.8 and p = 0.9
- Probabilities of the output symbols:  $q_0$  = ? And  $q_1$  = ?

$$q_0 = pP + \overline{p}\overline{P}$$

$$q_1 = p\overline{P} + \overline{p}P$$

• The backward probabilities:  $Q_{00}$ ,  $Q_{01}$ ,  $Q_{10}$ ,  $Q_{11}$  = ?

$$Q_{00} = \frac{p_0 P_{00}}{q_0}$$

$$Q_{10} = \frac{p_1 P_{10}}{q_0}$$

$$Q_{01} = \frac{p_0 P_{01}}{q_1}$$

$$Q_{11} = \frac{p_1 P_{11}}{q_1}$$

- Example 4.5 (Cont.)
  - Necessary and sufficient conditions on p and P for  $Q_{00}>Q_{10}$  and  $Q_{01}>Q_{11}$

$$Q_{00} = \frac{p_0 P_{00}}{q_0}$$
  $Q_{10} = \frac{p_1 P_{10}}{q_0}$   $Q_{01} = \frac{p_0 P_{01}}{q_1}$   $Q_{11} = \frac{p_1 P_{11}}{q_1}$ 

#### 4.3 System Entropies

- The input A and the output B of a channel  $\Gamma$ 
  - the input entropy  $H(\mathcal{A}) = \sum_i p_i \log \frac{1}{p_i}$
  - the output entropy  $H(\mathcal{B}) = \sum_{j} q_{j} \log \frac{1}{q_{j}}$
  - If  $b=b_j$  is received, then uncertainty about A is the conditional entropy

$$H(\mathcal{A} \mid b_j) = \sum_{i} \Pr(a_i \mid b_j) \log \frac{1}{\Pr(a_i \mid b_j)} = \sum_{i} Q_{ij} \log \frac{1}{Q_{ij}}$$

• the equivocation of A with respect to B

$$H(\mathcal{A} \mid \mathcal{B}) = \sum_{j} q_{j} H(\mathcal{A} \mid b_{j}) = \sum_{j} q_{j} \left( \sum_{i} Q_{ij} \log \frac{1}{Q_{ij}} \right) = \sum_{i} \sum_{j} R_{ij} \log \frac{1}{Q_{ij}}$$

#### System Entropies (Cont.)

• Similarly, if  $a_i$  is sent, then the uncertainty about B is the conditional entropy

$$H(\mathcal{B} \mid a_i) = \sum_{j} \Pr(b_j \mid a_i) \log \frac{1}{\Pr(b_j \mid a_i)} = \sum_{j} P_{ij} \log \frac{1}{P_{ij}}$$

the equivocation of B with respect to A

$$H(\mathcal{B} \mid \mathcal{A}) = \sum_{i} p_i H(\mathcal{B} \mid a_i) = \sum_{i} p_i \left( \sum_{j} P_{ij} \log \frac{1}{P_{ij}} \right) = \sum_{i} \sum_{j} R_{ij} \log \frac{1}{P_{ij}}$$

the joint entropy

$$H(\mathcal{A}, \mathcal{B}) = \sum_{i} \sum_{j} \Pr(a_i, b_j) \log \frac{1}{\Pr(a_i, b_j)} = \sum_{i} \sum_{j} R_{ij} \log \frac{1}{R_{ij}}$$

#### System Entropies (Cont.)

• If A and B are statistically independent, i.e.  $R_{ij}=p_iq_j$ , then

$$H(\mathcal{A}, \mathcal{B}) = H(\mathcal{A}) + H(\mathcal{B}) \tag{4.5}$$

 In general, A and B are related, rather than independent, then

$$H(\mathcal{A}, \mathcal{B}) = H(\mathcal{A}) + H(\mathcal{B} \mid \mathcal{A}) \tag{4.6}$$

$$H(\mathcal{A}, \mathcal{B}) = H(\mathcal{B}) + H(\mathcal{A} \mid \mathcal{B}) \tag{4.7}$$

• We call H(A), H(B), H(A|B), H(B|A), and H(A,B) the system entropies.

## 4.4 System Entropies for the Binary Symmetric Channel

The input and output entropies for BSC are

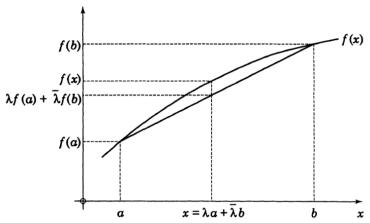
$$H(\mathcal{A}) = -p \log p - \overline{p} \log \overline{p} = H(p),$$
  $H(\mathcal{B}) = -q \log q - \overline{q} \log \overline{q} = H(q),$  where  $q = pP + \overline{p}\overline{P}.$ 

• Definition: A function  $f: [0,1] \to R$  is strictly convex, if for  $a,b \in [0,1]$  and  $x = \lambda a + \bar{\lambda} b$  with  $0 \le \lambda \le 1$ ,  $f(x) \ge \lambda f(a) + \bar{\lambda} f(b)$ ,

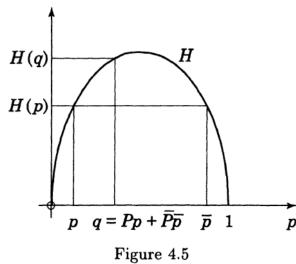
with equality if and only if x = a or b, that is, a = b or  $\lambda = 0$  or 1.

#### System Entropies for BSC (Cont.)

- Lemma 4.6
  - If a function  $f:[0,1] \to R$  is continuous on the interval [0,1] and twice differentiable on (0,1), with f''(x) < 0 for all  $x \in (0,1)$ , then f is strictly convex.
- Corollary 4.7
  - The entropy function H(p) is strictly convex on [0,1].



$$f(\lambda a + \bar{\lambda}b) \ge \lambda f(a) + \bar{\lambda}f(b)$$



$$H(pP + \bar{p}\bar{P}) \ge PH(p) + \bar{P}H(\bar{p})$$

$$H(pP + \bar{p}\bar{P}) \ge pH(P) + \bar{p}H(\bar{P})$$

### System Entropies for BSC (Cont.)

The BSC satisfies

$$H(\mathcal{B}) \ge H(\mathcal{A}), \quad (4.8)$$

with equality if and only if p = 1/2 or the channel is totally unreliable (P = 0) or reliable (P = 1)

Transmission through the BSC generally increases uncertainty

Note in BSC, 
$$q = pP + \bar{p}\bar{P}$$

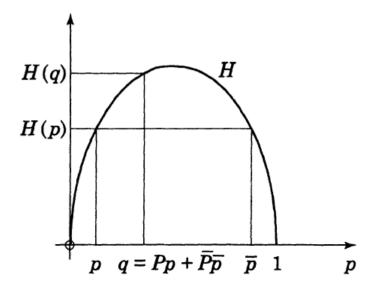


Figure 4.5

$$H(pP + \bar{p}\bar{P}) \ge PH(p) + \bar{P}H(\bar{p})$$