

Def. 4.9.

Suppose a source produces R bits per second (rate R).

Hence, NR bits in N seconds.

Total no. of messages in N seconds: 2^{NR} (ass. integer)

M codewords available for encoding all messages

$$M = 2^{NR} \Leftrightarrow R = \frac{\log M}{N}$$

(No. of bits per channel use.)

La. 4.13.

$(\underline{X}_N, \underline{Y}_N)$ is a DMC iff $\forall l=1, \dots, N$

$$\begin{aligned} P(Y_l = b_l \mid X_1 = a_1, \dots, X_N = a_N, Y_1 = b_1, \dots, Y_{l-1} = b_{l-1}) \\ = P(Y_l = b_l \mid X_l = a_l) \end{aligned}$$

Proof. " \Leftarrow "

$$P(\underline{Y}_N = \underline{b}_N \mid \underline{X}_N = \underline{a}_N)$$

$$= P(\underline{Y}_N = \underline{b}_N \mid \underline{X}_N = \underline{a}_N, \underline{Y}_{N-1} = \underline{b}_{N-1}) \cdot \frac{P(\underline{Y}_{N-1} = \underline{b}_{N-1} \mid \underline{X}_{N-1} = \underline{a}_{N-1})}{P(\underline{X}_N = \underline{a}_N)}$$

$$\stackrel{(\text{Ass.})}{=} P(Y_N = b_N \mid X_N = a_N) \cdot P(\underline{Y}_{N-1} = \underline{b}_{N-1} \mid \underline{X}_{N-1} = \underline{a}_{N-1})$$

$$\begin{aligned} &= P(Y_1 = b_1 \mid X_1 = a_1) \cdot P(Y_2 = b_2 \mid X_2 = a_2) \cdot \dots \cdot P(Y_N = b_N \mid X_N = a_N) \\ &= \dots = \prod_{i=1}^N P(Y_i = b_i \mid X_i = a_i) \end{aligned}$$

\Rightarrow

$$P(Y_e = b_e | X_N = a_N, Y_{e-1} = b_{e-1})$$

$$= \frac{P(Y_e = b_e | X_N = a_N)}{P(Y_{e-1} = b_{e-1} | X_N = a_N)}$$

$$\neq \frac{\sum_{b_{e-1} \in \mathcal{Y}} P(Y_N = b_N | X_N = a_N)}{P(Y_N = b_N | X_N = a_N)}$$

Ex. $= \frac{\sum_{b_{e-1} \in \mathcal{Y}} P(Y_N = b_N | X_N = a_N)}{P(Y_N = b_N | X_N = a_N)}$

Ass. $= P(Y_1 = b_e | X_1 = a_e)$

□

$\{(X_n, Y_n)\}$ is a sequence of independent r.v.s
then (X_N, Y_N) forms a DMC.

Th. 4.14. (Outline of the proof)

Use random coding, i.e., r.v.

$C_1, \dots, C_M \in \mathcal{X}^N$, $C_i = (C_{i1}, \dots, C_{iN})$, $i=1, \dots, M$
with $C_{ij} \in \mathcal{X}$, i.i.d. $\sim p(x)$, $i=1, \dots, M$, $j=1, \dots, N$.

Th. A. For a DMC with ML-decoding it holds
for all δ $0 \leq \delta \leq 1$, $j=1, \dots, M$

$$E(e_j(C_1, \dots, C_M)) \leq (M-1)^\delta \left(\sum_{j=1}^M \left(\sum_{i=1}^M p_i p_1(y_j | x_i)^{\frac{1}{1+\delta}} \right)^{1+\delta} \right)^N \quad \perp$$

$$\text{Set } G(\delta, p) = -\ln \left[\sum_{j=1}^M \left(\sum_{i=1}^M p_i p_1(y_j | x_i)^{\frac{1}{1+\delta}} \right)^{1+\delta} \right]$$

$$\text{and } R = \frac{\ln M}{N}.$$

$$E(e_j(C_1, \dots, C_M)) \leq \exp(-N(G(\delta, p) - \delta R))$$

$$\text{Set } G^*(R) = \max_{0 \leq \delta \leq 1} \max_p \{G(\delta, p) - \delta R\}$$

Th. B. For a DMC with ML decoding there
exists a code $C_1, \dots, C_M \in \mathcal{X}^N$ s.t.

$$\hat{e}(C_1, \dots, C_M) \leq 4e^{-NG^*(R)} \quad \perp$$

Proof. Use $2M$ ^{randomly} codewords. Then

$$\frac{1}{2M} \sum_{j=1}^{2M} E(e_j(c_1, \dots, c_{2M})) \leq e^{-NG^*\left(\frac{\ln 2M}{N}\right)}$$

There exists a sample c_1, \dots, c_{2M} s.t.

$$\frac{1}{2M} \sum_{j=1}^{2M} e_j(c_1, \dots, c_{2M}) \leq e^{-NG^*\left(\frac{\ln 2M}{N}\right)} \quad (*)$$

Remove M codewords, particularly with

$$e_R(c_1, \dots, c_{2M}) \geq 2 e^{-NG^*\left(\frac{\ln 2M}{N}\right)}$$

There are at most M , otherwise (*) would be violated.

For the remaining ones

$$e_j(c_{i_1}, \dots, c_{i_M}) \leq 4 e^{-NG^*\left(\frac{\ln 2M}{N}\right)} \quad \forall j=1, \dots, M. \quad \square$$

Th.C. If $R = \frac{\ln M}{N} < C$, then

$$\begin{aligned} G^*(R) &= \max_p \max_{0 \leq \gamma \leq 1} \{ G(\gamma, p) - \gamma R \} \\ &\geq \max_{0 \leq \gamma \leq 1} \{ G(\gamma, p^*) - \gamma R \} > 0 \end{aligned}$$

where p^* denotes the capacity-achieving distr.

(Ref. RM, p. 103-114)

5. Rate Distortion Theory

Motivation:

- a) By the source coding theorem (Th. 3.7 & 3.9):
error free / lossless encoding needs at least
on average $H(X)$ bits per symbol.
What can be said if fewer bits are available.
- b) Signal is represented by bits. What is the
min. no. of bits needed not to exceed
a certain max. distortion.

Example 5.1.

- a) Representing a real number by K bits:

$$\mathcal{X} = \mathbb{R}, \quad \mathcal{X}^1 = \{(b_1, \dots, b_K) \mid b_i \in \{0, 1\}\}$$

- b) 1-bit quantization

$$\mathcal{X} = \mathbb{R}, \quad \mathcal{X}^1 = \{0, 1\}$$

- c) Representing a 28×28 gray-scale picture (8 values)
by R bits

$$\mathcal{X} = 2^{28 \cdot 28 \cdot 3}, \quad \mathcal{X}^1 = \{1, 2, \dots, 2^R\}$$

\mathcal{X} is called source alphabet, \mathcal{X}^1 reproduction alph.

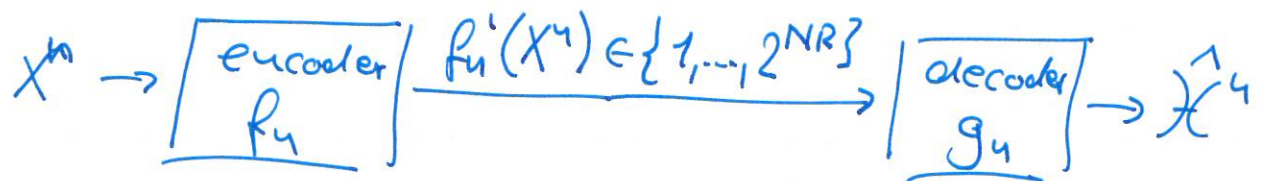
Both are assumed to be finite.

General situation:

X_1, \dots, X_n i.i.d. r.v. $\sim p(x)$, $x \in \mathcal{X}$ output of some source.

$f_n(X^n)$ encoding of X^n by an index $1, 2, \dots, 2^{NR}$.

$g_n: \{1, \dots, 2^{NR}\} \rightarrow \hat{\mathcal{X}}^n$ decoding by a reproduction in $\hat{\mathcal{X}}^n$.



Def. 5.2. A distortion function/measure is a mapping $d: \mathcal{X} \times \hat{\mathcal{X}} \rightarrow \mathbb{R}_+$.

Examples.

a) Hamming distance, $\mathcal{X} = \hat{\mathcal{X}} = \{0, 1\}$

$$d(x, \hat{x}) = \begin{cases} 0, & x = \hat{x} \\ 1, & \text{otherwise} \end{cases}$$

b) Squared error: $d(\hat{x}, x) = (x - \hat{x})^2$

Def. 5.3. The distortion measure between sequences x^n, \hat{x}^n is defined

$$d(x^n, \hat{x}^n) = \frac{1}{n} \sum_{i=1}^n d(x_i, \hat{x}_i).$$