# Digital Communication Laboratory (Digital Comm Lab)

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#### Reference

• John G. Proakis, Masoud Salehi, and Gerhard Bauch. "Contemporary communication systems using MATLAB". Third Edition, Cengage Learning, 2012.

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#### Software

• MATLAB R2017b

### 1. Signals and Linear Systems

• ILLUSTRATIVE PROBLEM 1.1

 $x(t) = A\Pi\left(\frac{t}{2t_0}\right) = \begin{cases} A, & |t| < t_0\\ \frac{A}{2}, & t = \pm t_0\\ 0, & \text{otherwise} \end{cases}$ (1.2.19)for  $|t| \le T_0/2$ , where  $t_0 < T_0/2$ .  $\Pi(t) = \begin{cases} 1, & |t| < \frac{1}{2} \\ \frac{1}{2}, & t = \pm \frac{1}{2} \\ 0, & \text{otherwise} \end{cases}$ (1.2.20)x(t)Α  $\frac{T_0}{2}$  $-\frac{T_0}{2}$  $-T_0$  $T_0$  $-t_0$  $t_0$ 

• Huffman Coding:

#### ILLUSTRATIVE PROBLEM

**Illustrative Problem 4.1** [Huffman Coding] Design a Huffman code for a source with alphabet  $\mathscr{X} = \{x_1, x_2, ..., x_9\}$  and corresponding probability vector

p = (0.2, 0.15, 0.13, 0.12, 0.1, 0.09, 0.08, 0.07, 0.06)

Find the average codeword length of the resulting code and compare it with the entropy of the source.

$$H(X) = -\sum_{x \in \mathscr{X}} p(x) \log p(x)$$

#### ILLUSTRATIVE PROBLEM

**Illustrative Problem 4.2** [Huffman Coding] A discrete-memoryless information source with alphabet

$$\mathscr{X} = \{x_1, x_2, \dots, x_6\}$$

and the corresponding probabilities

 $p = \{0.1, 0.3, 0.05, 0.09, 0.21, 0.25\}$ 

is to be encoded using Huffman coding.

- 1. Determine the entropy of the source.
- Find a Huffman code for the source and determine the efficiency of the Huffman code.
- 3. Now design a Huffman code for source sequences of length 2 and compare the efficiency of this code with the efficiency of the code derived in part 2.

- 4.3. Quantization
  - Scalar Quantization
    - Uniform Quantization
    - Nonuniform Quantization
  - Vector Quantization

- 4.3. Quantization
  - Scalar Quantization
    - Uniform Quantization
    - Nonuniform Quantization: Many physical signals, such as speech signals, have the characteristic that small signal amplitudes occur more frequently than large signal amplitudes. However, a uniform quantizer provides the same spacing between successive levels throughout the entire dynamic range of the signal. A better approach would be to have a nonuniform quantizer, which provides more closely spaced levels at the small signal amplitudes and more widely spaced levels at the large signal amplitudes.
    - A nonuniform quantizer characteristic is usually obtained by passing the signal through a nonlinear device that compresses the signal amplitudes, followed by a uniform PCM quantizer.
    - Compressor + Expander = Compander

- 4.3. Quantization
  - Nonuniform Quantization:



- Source Coding
  - Compand a signal
    - Compand
      - 'mu/compressor'

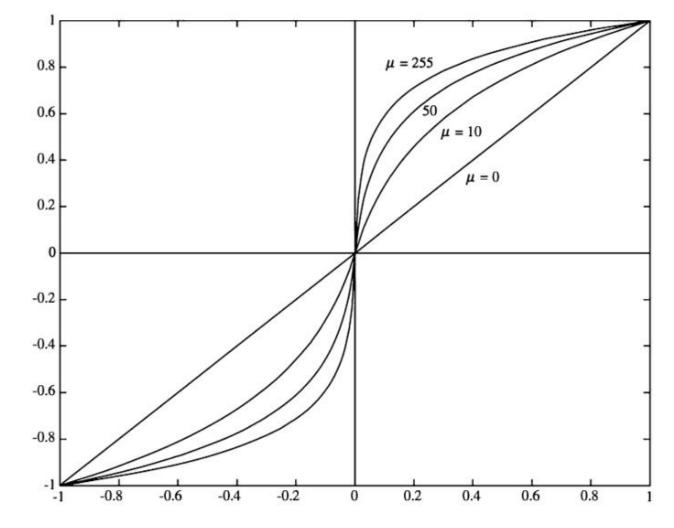


Figure 4.14: The  $\mu$ -law compander

- 4.3.3. Pulse Code Modulation
  - PCM
    - Sampling at a rate higher than Nyquist rate
    - Quantization
      - Uniform PCM
      - Nonuniform PCM

- 4.3.4. Differential Pulse Code Modulation (DPCM)
  - Samples are usually correlated random variables
  - In the simplest form of DPCM, difference between two adjacent samples is quantized.

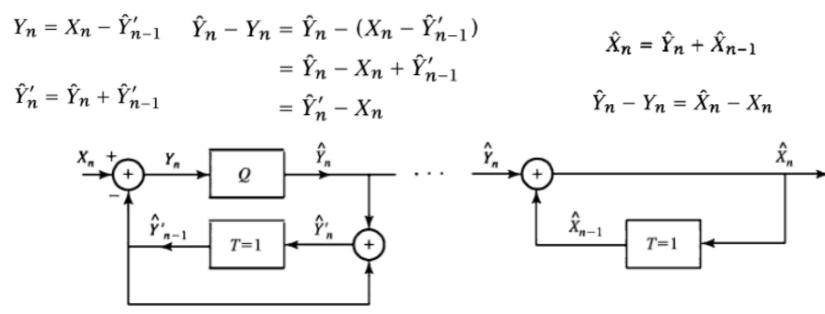


Figure 4.18: A simple DPCM encoder and decoder

- 4.3.4. Differential Pulse Code Modulation (DPCM)
  - MATLAB
    - Source Coding
      - DPCM
    - Predictor
    - dpcmenco
    - dpcmdeco
    - dpcmopt

• Binary Signal Transmission:

$0 \rightarrow s_0(t)$ ,	$0 \leq t \leq T_b$
$1 \rightarrow s_1(t)$ ,	$0 \leq t \leq T_b$

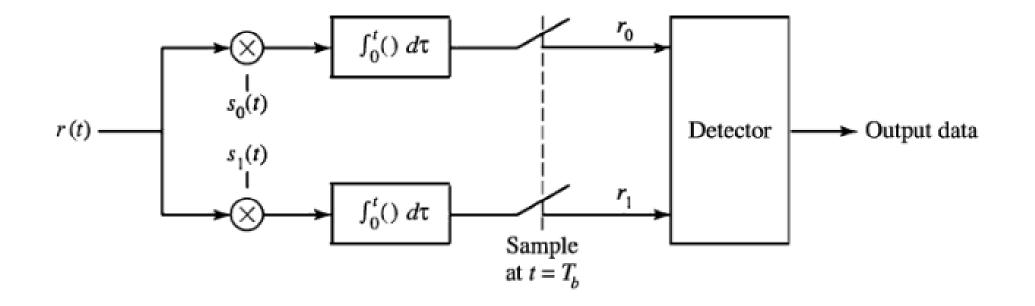
• Additive White Gaussian Noise (AWGN):

 $r(t) = s_i(t) + n(t), \quad i = 0, 1, \quad 0 \le t \le T_b$ 

- Optimum Receiver for AWGN Channel:
  - Signal Correlator
  - Matched Filter
  - Detector

- Optimum Receiver for AWGN Channel:
  - Signal Correlator

$$r_0(t) = \int_0^t r(\tau) s_0(\tau) d\tau$$
$$r_1(t) = \int_0^t r(\tau) s_1(\tau) d\tau$$



- Optimum Receiver for AWGN Channel:
  - Signal Correlator
  - Example 1:

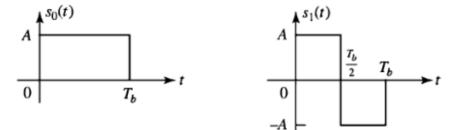


Figure 5.2: Signal waveforms  $s_0(t)$  and  $s_1(t)$  for a binary communication system

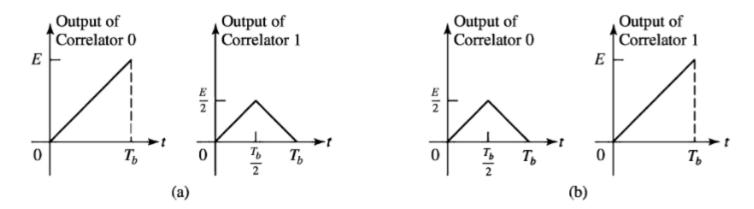


Figure 5.3: Noise-free correlator outputs. (a)  $s_0(t)$  was transmitted. (b)  $s_1(t)$  was transmitted

#### • Optimum Receiver for AWGN Channel:

- Signal Correlator
- Example 1:

$$r_{0} = \int_{0}^{T_{b}} r(t)s_{0}(t) dt \qquad r_{1} = \int_{0}^{T_{b}} r(t)s_{1}(t) dt$$
$$= \int_{0}^{T_{b}} s_{0}^{2}(t) dt + \int_{0}^{T_{b}} n(t)s_{0}(t) dt \qquad = \int_{0}^{T_{b}} s_{0}(t)s_{1}(t) dt + \int_{0}^{T_{b}} n(t)s_{1}(t) dt$$
$$= n_{1}$$

$$n_0 = \int_0^{T_b} n(t)s_0(t) dt \qquad p(r_0 \mid s_0(t) \text{ was transmitted}) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(r_0 - E)^2/2\sigma^2}$$
$$n_1 = \int_0^{T_b} n(t)s_1(t) dt \qquad p(r_1 \mid s_0(t) \text{ was transmitted}) = \frac{1}{\sqrt{2\pi}\sigma} e^{-r_1^2/2\sigma^2}$$

- Optimum Receiver for AWGN Channel:
  - Signal Correlator
  - Example 1:

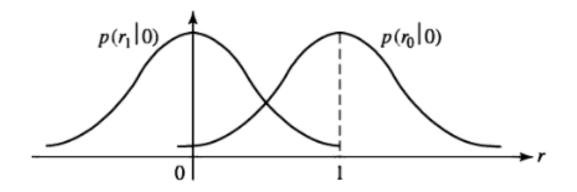


Figure 5.4: Probability density functions  $p(r_0 | 0)$  and  $p(r_1 | 0)$  when  $s_0(t)$  is transmitted

#### • Optimum Receiver for AWGN Channel:

- Signal Correlator
- Example 2: Illustrative Problem 5.2 [Correlation of Signal Waveforms] Sample the signal waveforms in Illustrative Problem 5.1 at a rate  $F_s = 20/T_b$  (sampling interval  $T_s = T_b/20$ ) and perform the correlation of r(t) with  $s_0(t)$  and  $s_1(t)$  numerically; that is, compute and plot

$$r_0(kT_s) = \sum_{n=1}^k r(nT_s)s_0(nT_s), \quad k = 1, 2, ..., 20$$

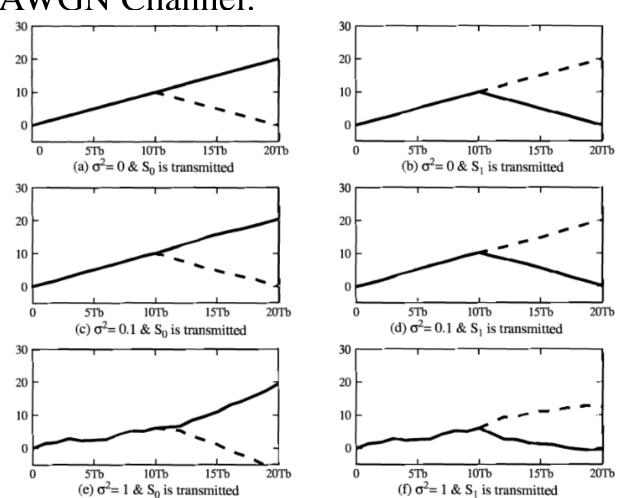
and

$$r_1(kT_s) = \sum_{n=1}^k r(nT_s)s_1(nT_s), \quad k = 1, 2, \dots, 20$$

when (a)  $s_0(t)$  is transmitted signal and (b)  $s_1(t)$  is the transmitted signal.

Repeat the above computations and plots when the signal samples  $r(kT_s)$  are corrupted by additive white Gaussian noise samples  $n(kT_s)$ ,  $1 \le k \le 20$ , which have zero mean and variance  $\sigma^2 = 0.1$  and  $\sigma^2 = 1$ .

- Optimum Receiver for AWGN Channel:
  - Signal Correlator
  - Example 2:



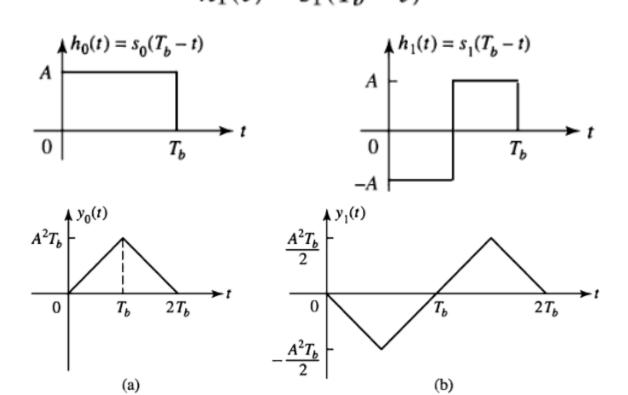
#### • Optimum Receiver for AWGN Channel:

• Matched Filter

$$h(t) = s(T_b - t), \qquad 0 \le t \le T_b$$
$$y(t) = \int_0^t s(\tau)h(t - \tau) d\tau$$
$$y(t) = \int_0^t s(\tau)s(T_b - t + \tau) d\tau$$
$$y(T_b) = \int_0^{T_b} s^2(t) dt = E$$

- Optimum Receiver for AWGN Channel:
  - Matched Filter
  - Example 1:

 $h_0(t) = s_0(T_b - t)$  $h_1(t) = s_1(T_b - t)$ 



#### • Optimum Receiver for AWGN Channel:

- Matched Filter
- Example 2:

**Illustrative Problem 5.4 [Match Filtering of Signal Waveforms]** Sample the signal waveform in Illustrative Problem 5.3 at a rate of  $F_s = 20/T_b$  and perform the matched filtering of the received signal r(t) with  $s_0(t)$  and  $s_1(t)$  numerically; that is, compute and plot

$$y_0(kT_s) = \sum_{n=1}^k r(nT_s)s_0(kT_s - nT_s), \quad k = 1, 2, ..., 20$$

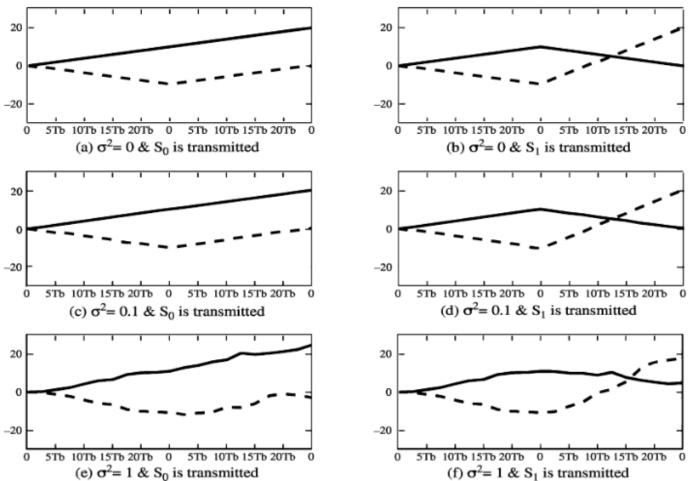
and

$$y_1(kT_s) = \sum_{n=1}^k r(nT_s)s_1(kT_s - nT_s), \quad k = 1, 2, ..., 20$$

when (a)  $s_0(t)$  is the transmitted signal and (b)  $s_1(t)$  is the transmitted signal.

Repeat the above computations when the signal samples  $r(kT_s)$  are corrupted by additive white Gaussian noise samples  $n(kT_s)$ ,  $1 \le k \le 20$ , which have zero mean and variance  $\sigma^2 = 0.1$  and  $\sigma^2 = 1$ .

- Optimum Receiver for AWGN Channel:
  - Matched Filter
  - Example 2:

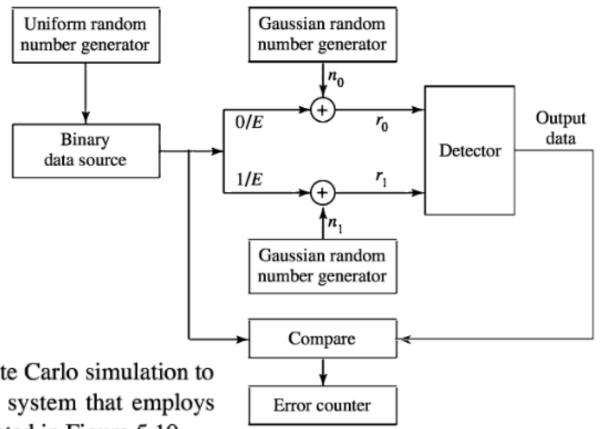


• Probability of error vs. signal-to-noise ratio

$$P_e = \frac{1}{\sqrt{2\pi}} \int_E^\infty e^{-x^2/2\sigma_x^2} dx$$
$$= \frac{1}{\sqrt{2\pi}} \int_{\sqrt{E/N_0}}^\infty e^{-x^2/2} dx$$
$$= Q\left(\sqrt{\frac{E}{N_0}}\right)$$

- Qfunc
- semilogy

• Monte Carlo Simulation of Binary Communication System



**Illustrative Problem 5.6 [Monte Carlo Simulation]** Use Monte Carlo simulation to estimate and plot  $P_e$  versus SNR for a binary communication system that employs correlators or matched filters. The model of the system is illustrated in Figure 5.10.

• Signal Constellation Diagrams for Binary Signals

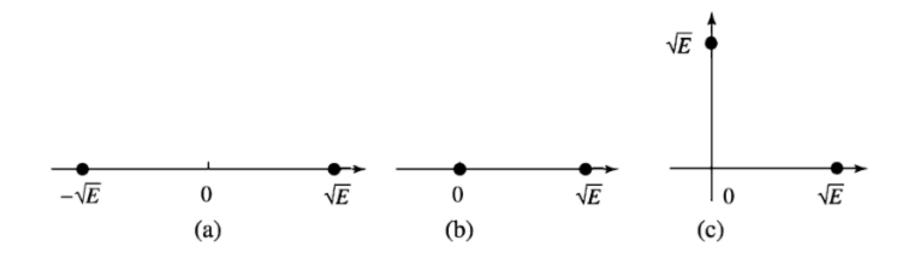


Figure 5.21: Signal constellations for binary signals. (a) Antipodal signals. (b) On–off signals. (c) Orthogonal signals

• Signal Constellation Diagrams for Binary Signals

**Illustrative Problem 5.11 [Noise Effect on the Constellation]** The effect of noise on the performance of a binary communication system can be observed from the received signal plus noise at the input to the detector. For example, let us consider binary orthogonal signals, for which the input to the detector consists of the pair of random variables  $(r_0, r_1)$ , where either

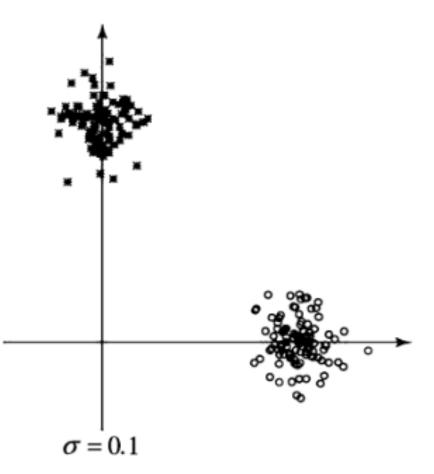
$$(r_0, r_1) = (\sqrt{E} + n_0, n_1)$$

or

$$(r_0, r_1) = (n_0, \sqrt{E} + n_1)$$

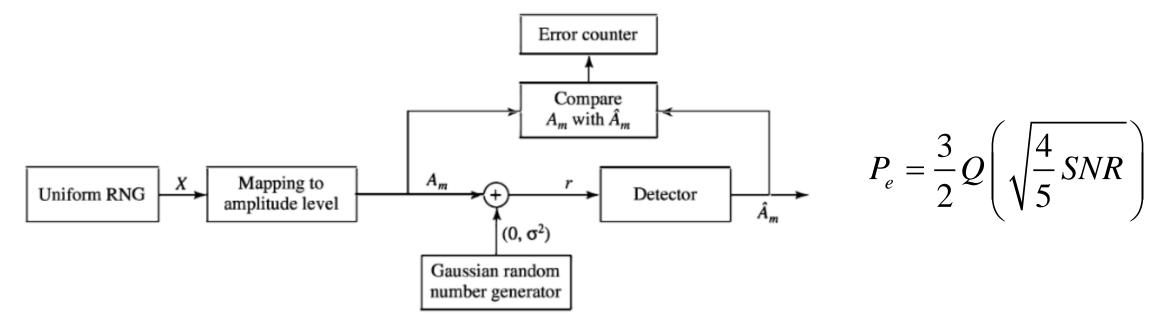
The noise random variables  $n_0$  and  $n_1$  are zero-mean, independent Gaussian random variables with variance  $\sigma^2$ . As in Illustrative Problem 5.6, use Monte Carlo simulation to generate 100 samples of  $(r_0, r_1)$  for each value of  $\sigma = 0.1$ ,  $\sigma = 0.3$ , and  $\sigma = 0.5$ , and plot these 100 samples for each  $\sigma$  on different two-dimensional plots. The energy *E* of the signal may be normalized to unity.

• Signal Constellation Diagrams for Binary Signals

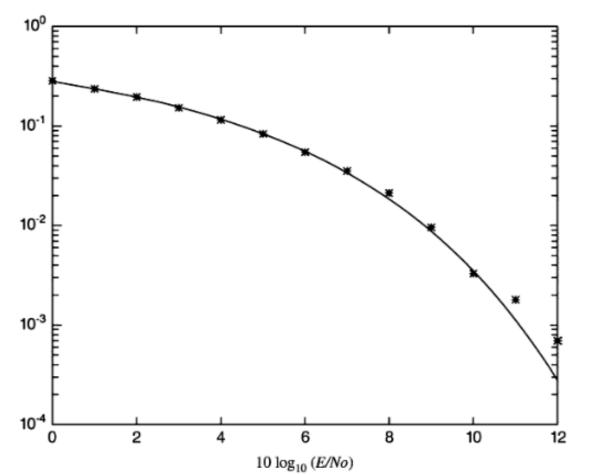


#### • Multiamplitude Signal Simulation

**Illustrative Problem 5.12 [Multiamplitude Signal Simulation]** Perform a Monte Carlo simulation of the four-level (quaternary) PAM communication system that employs a signal correlator, as described earlier, followed by an amplitude detector. The model for the system to be simulated is shown in Figure 5.26.



• Multiamplitude Signal Simulation



- MATLAB Tool
  - bertool
  - doc\_bpsk

