

### 8.3 Digital PAM with Noise

**8.13.** In this section, the transmitted signal is still in the form of digital PAM as in the previous subsection:

$$x(t) = \sum_n m[n]p(t - nT_s)$$

However, here, we also consider the effect of additive noise. Therefore, the received signal is

$$y(t) = x(t) + N(t)$$

where  $N(t)$  is a random noise process.

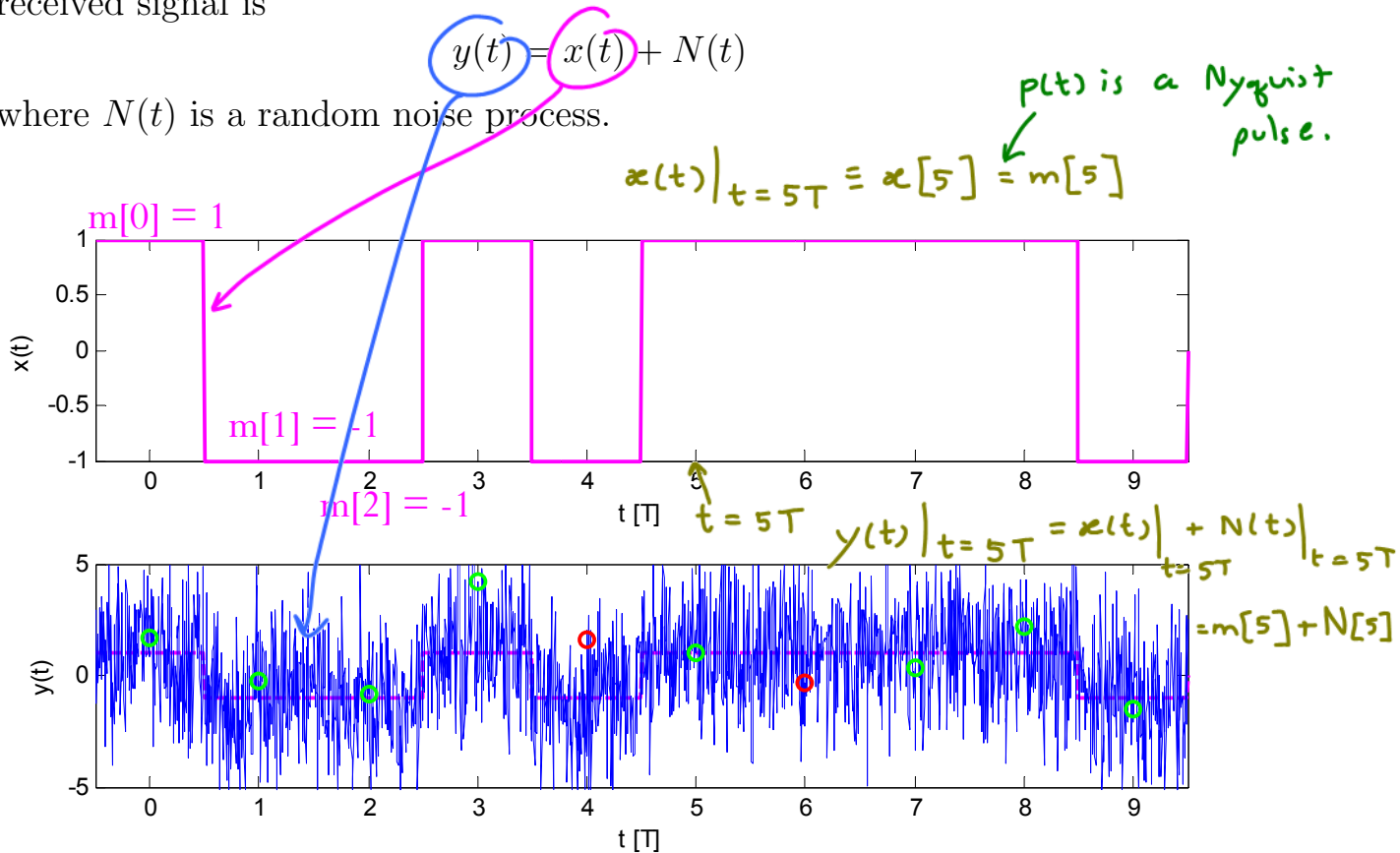


Figure 47: Digital PAM with Noise

**8.14.** Note that

- The noise  $N(t)$  is random.
- The message  $m[n]$  should be random (at least from the perspective of the receiver; if the receiver had known in advance the value of  $m[n]$ , there would have been no point of transmitting  $m[n]$ ).

o This makes  $x(t)$  and  $y(t)$  random.

To emphasize the randomness in the signals under consideration, we sometimes write  $M[n]$ ,  $X(t)$ , and  $Y(t)$  using capital letters<sup>26</sup>.

**8.15. Simple receiver:** For simplicity, let's assume that our receiver simply samples  $y(t)$  every  $T$  seconds to get

Represented by circles in the Figure  $\leftarrow y[n] = y(t)|_{t=nT} = (x(t) + N(t))|_{t=nT} = m[n] + N(t)|_{t=nT}$

*p(t) is a Nyquist pulse*

When the alphabet  $\mathcal{A}$  contains only two symbols of opposite sign ( $\mathcal{A} = \{-a, a\}$ , where  $a > 0$ ), the decoded value  $\hat{m}[n]$  of our  $m[n]$  can be found by



$$\hat{m}[n] = \begin{cases} a, & y[n] \geq 0 \\ -a, & y[n] < 0 \end{cases}$$

*y[n] is closer to "a"* (circled in blue)  
*y[n] is closer to "-a"* (circled in blue)

Here we use "0" as the **threshold** level/value. Turn out that this middle point is the *optimal* threshold to use when the two possible symbol values from the alphabet are equally likely.

**Example 8.16.** In Figure 47,  $\mathcal{A} = \{-1, 1\}$ .

*minimum distance decoding*  
*Idea: noise is usually small.*

$n$	0	1	2	3	4	5	6	7	8	9
$m[n]$	1	-1	-1	1	-1	1	1	1	1	-1
$y[n]$	1.69	-0.27	-0.81	4.20	1.58	1.04	-0.34	0.35	2.19	-1.51
$\hat{m}[n]$	1	-1	-1	1	1	1	-1	1	1	-1

*error* (circled in red around the 1 and -1 in the last row)

Note that the value of the noise  $N(t)$  at  $t = 4T$  is too positive. Even when "-1" was transmitted, the received value is 1.58 which exceeds 0. Therefore, we get an error at  $n = 4$ .

Similarly that the value of the noise  $N(t)$  at  $t = 6T$  is too negative. Even when "1" was transmitted, the received value is -0.34 which is lower than 0. Therefore, we get an error at  $n = 6$ .

Among the **ten symbols** sent in this example, there are **two symbol errors**. Therefore, the **symbol error rate (SER)** or symbol error probability is **2/10**.

Because there are two symbols in the alphabet, **each symbol** transmission conveys **1 bit**. Hence, the **bit error rate (BER)** or bit error probability is also **2/10**.

<sup>26</sup>Caution: Here, capital letters represent random variables/processes. In earlier sections, we used capital letter to represent Fourier transform. However, we won't talk about Fourier transform here; so confusion can be avoided.

(AWGN)

Normal

**8.17. Additive White Gaussian Noise:** At each time instant  $t$ , the noise  $N(t)$  is usually modeled by a Gaussian random variable with mean 0 and standard deviation  $\sigma_N$ ,

pdf

$$f_{N(t)}(n) = \frac{1}{\sqrt{2\pi}\sigma_N} e^{-\frac{1}{2}\left(\frac{n}{\sigma_N}\right)^2}.$$

Furthermore, the “white” part means that the noise values at different time instants are **independent**.

**Definition 8.18.** In general, a **Gaussian (normal) random variable**  $X$  with mean  $m$  and standard deviation  $\sigma$  is characterized by its probability density function (PDF):

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-m}{\sigma}\right)^2}.$$

To talk about such  $X$ , we usually write  $X \sim \mathcal{N}(m, \sigma^2)$ . Probability involving  $X$  can be evaluated by

$$P[X \in A] = \int_A f_X(x) dx.$$

In particular,

$$P[X \in [a, b]] = \int_a^b f_X(x) dx = F_X(b) - F_X(a)$$

where  $F_X(x) = \int_{-\infty}^x f_X(t) dt$  is called the cumulative distribution function (CDF) of  $X$ .

We usually express probability involving Gaussian random variable via the  $Q$  function which is defined by

qfunc

$$Q(z) = \int_z^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx.$$

Note that  $Q(z)$  is the same as  $P[Z > z]$  where  $Z \sim \mathcal{N}(0, 1)$ ; that is  $Q(z)$  is the probability of the “tail” of  $\mathcal{N}(0, 1)$ .

It can be shown that

- $Q$  is a decreasing function

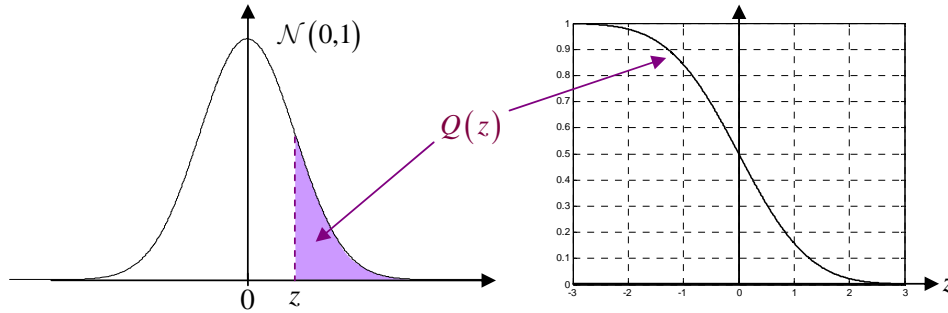
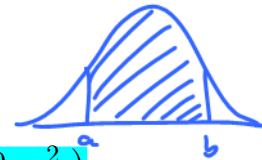


Figure 48:  $Q$ -function

- $Q(0) = \frac{1}{2}$
- $Q(-z) = 1 - Q(z)$ 
  - This is useful for converting the argument of the  $Q$  function to positive value.
- For  $X \sim \mathcal{N}(m, \sigma^2)$ ,

$$P[X > c] = Q\left(\frac{c - m}{\sigma}\right).$$



**8.19.** Three important noise probabilities for  $N \sim \mathcal{N}(0, \sigma_N^2)$ :

$$P[N > c] = Q\left(\frac{c}{\sigma_N}\right), \quad P[N < c] = 1 - Q\left(\frac{c}{\sigma_N}\right), \quad P[a < N < b] = P[N < b] - P[N < a]$$

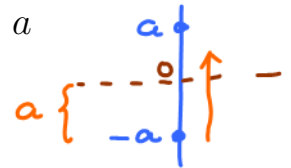
$$= P[N > a] - P[N > b] = Q\left(\frac{a}{\sigma_N}\right) - Q\left(\frac{b}{\sigma_N}\right)$$

Note that all strict inequalities above can also be replaced by the ones that also include equalities because the noise is a continuous random variable and hence including one particular noise value does not change probability.

**8.20.** For the simple receiver in 8.15, suppose  $N(t) \sim \mathcal{N}(0, \sigma^2)$ .

(a) When a “ $-a$ ” was transmitted, error occurs when  $N(t) > a$

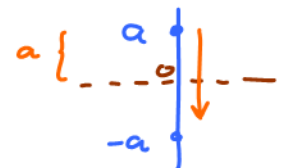
$$P[N(t) > a] = Q\left(\frac{a}{\sigma}\right)$$



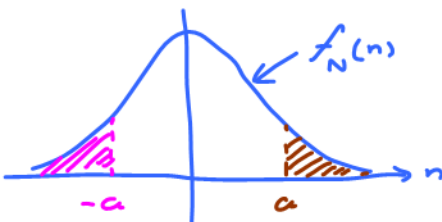
(b) When a “ $+a$ ” was transmitted, error occurs when  $N(t) < -a$

$$P[N(t) < -a] = 1 - P[N(t) \geq -a]$$

$$= 1 - Q\left(\frac{-a}{\sigma}\right)$$



$$= 1 - (1 - Q\left(\frac{a}{\sigma}\right)) = Q\left(\frac{a}{\sigma}\right)$$



A formula that connects these events with the (combined) error probability is called the **Total Probability Theorem**: If a (finite or infinitely) countable collection of events  $\{B_1, B_2, \dots\}$  is a partition of  $\Omega$ , then

$$P(A) = \sum_i P(A|B_i)P(B_i). \tag{62}$$

In particular,

(decoding) error event  
 $\mathcal{E}$

$$P(A) = P(A|B)P(B) + P(A|B^c)P(B^c).$$

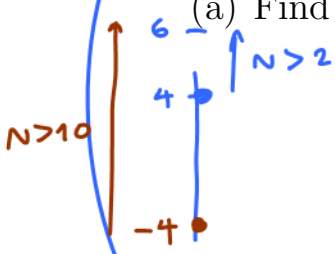
Here, we replace event  $A$  by the error event  $\mathcal{E}$ . Event  $B$  is defined to be the event that the transmitted symbol is "a". The error probability is then

$$P(\mathcal{E}) = \underbrace{P(\mathcal{E}|B)}_{Q(\frac{a}{\sigma})} P(B) + \underbrace{P(\mathcal{E}|B^c)}_{Q(\frac{a}{\sigma})} P(B^c) = Q(\frac{a}{\sigma}) (P(B) + P(B^c))$$

~~1~~

**Example 8.21.** In a digital PAM system, equally-likely symbols are selected from an alphabet set  $\mathcal{A} = \{-a, a\}$ . The pulse used in the transmitted signal is a Nyquist pulse. The additive noise at each particular time instant is Gaussian with mean 0 and standard deviation 2.

(a) Find the probability that the received signal at a particular time is  $> 6$ .



"a" was transmitted event A

$$P(A|B) = P[N > 2] = Q\left(\frac{2}{2}\right) = Q(1) \approx 0.1587$$

$$P(A|B^c) = P[N > 10] = Q\left(\frac{10}{2}\right) = Q(5) \approx 2.9 \times 10^{-7}$$

$$P(A) = P(A|B)P(B) + P(A|B^c)P(B^c) = \frac{Q(1) + Q(5)}{2} \approx 0.0793$$

(b) Find the symbol error probability when 0 is used as the threshold level for the decoding decision at the receiver.

$$P(\mathcal{E}) = Q\left(\frac{a}{\sigma}\right) = Q\left(\frac{4}{2}\right) = Q(2) \approx 0.0228$$

SNR = signal-to-noise (power) ratio =  $\frac{a^2}{\mathbb{E}[N^2(t)]} = \frac{a^2}{\sigma^2}$

$$P(\mathcal{E}) = Q(\sqrt{\text{SNR}})$$

$$\text{Var } X = \mathbb{E}[X^2] - (\mathbb{E}X)^2$$

$$\mathbb{E}[X^2] = \text{Var } X + (\mathbb{E}X)^2 = \sigma_X^2 + (\mathbb{E}X)^2$$