

# Communication Systems II

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# Outline

- Optimum decision rule
  - MAP criterion
  - ML criterion
  - Minimum Euclidean distance rule
  - Maximum correlation rule
  
- Average symbol rate
  - M-ary PAM
  - M-ary orthogonal signals

# Optimum Detector

## ■ Posteriori probability

$$P(\text{signal } \mathbf{s}_m \text{ was transmitted} | \mathbf{y})$$

## ■ Optimum decision rule

- Select the signal corresponding to the maximum set of posteriori probabilities:

choose  $m$  such that  $\{P(\mathbf{s}_m | \mathbf{y})\}_{m=1}^M$  is maximum

- which is called “*maximum a posteriori (MAP)*” criterion.

## ■ Bays' rule

$$P(\mathbf{s}_m | \mathbf{y}) = \frac{f(\mathbf{y} | \mathbf{s}_m)P(\mathbf{s}_m)}{f(\mathbf{y})}$$

- where  $P(\mathbf{s}_m)$  is called “a priori probability”.

## ■ MAP criterion

choose  $m$  such that  $\{P(\mathbf{s}_m|\mathbf{y})\}_{m=1}^M$  is maximum  
= choose  $m$  such that  $\left\{\frac{f(\mathbf{y}|\mathbf{s}_m)P(\mathbf{s}_m)}{f(\mathbf{y})}\right\}_{m=1}^M$  is maximum

- For equally probable case, that is,  $P(\mathbf{s}_m) = \frac{1}{M}$ , MAP criterion becomes

choose  $m$  such that  $\{f(\mathbf{y}|\mathbf{s}_m)\}_{m=1}^M$  is maximum

- ◆ which is called “**maximum likelihood (ML)**” criterion.

## ■ Definition

- likelihood function:  $f(\mathbf{y}|\mathbf{s}_m)$
- Log-likelihood function:  $\ln f(\mathbf{y}|\mathbf{s}_m)$

## ■ MAP criterion

choose  $m$  such that  $\{P(\mathbf{s}_m|\mathbf{y})\}_{m=1}^M$  is maximum

= choose  $m$  such that  $\left\{\frac{f(\mathbf{y}|\mathbf{s}_m)P(\mathbf{s}_m)}{f(\mathbf{y})}\right\}_{m=1}^M$  is maximum

for equally probable case => = choose  $m$  such that  $f(\mathbf{y}|\mathbf{s}_m)$

= choose  $m$  such that  $\ln f(\mathbf{y}|\mathbf{s}_m)$

## ■ 4-PAM case

### ● Likelihood function

$$f(y|s_m) = \frac{1}{\sqrt{\pi N_0}} e^{-(y-s_m)^2/N_0}, \quad m = 1, 2, \dots, M$$

### ● Log-Likelihood function

$$\ln f(y|s_m) = -\frac{1}{2} \log(\pi N_0) - \frac{(y - s_m)^2}{N_0}, \quad m = 1, 2, \dots, M$$

### ● ML criterion

$$\max_m \left[ -\frac{1}{2} \log(\pi N_0) - \frac{(y - s_m)^2}{N_0} \right], \quad m = 1, 2, \dots, M$$

$$= \max_m \left[ -\frac{(y - s_m)^2}{N_0} \right], \quad m = 1, 2, \dots, M$$

$$= \min_m [(y - s_m)^2], \quad m = 1, 2, \dots, M$$

$$= \min_m \|y - s_m\|, \quad m = 1, 2, \dots, M$$

- Generally, the output of the demodulator over AWGN channel can be written as

$$y_k = s_{mk} + n_k, \quad k = 1, 2, \dots, \underbrace{N}_{\text{dimension}}$$

- Its likelihood function is given as

$$f(y_k | s_{mk}) = \frac{1}{\sqrt{\pi N_0}} e^{-(y - s_{mk})^2 / N_0}, \quad m = 1, 2, \dots, M$$

- Joint likelihood function

$$f(\mathbf{y} | \mathbf{s}_m) = \frac{1}{(\pi N_0)^{N/2}} e^{-\sum_{k=1}^N (y - s_{mk})^2 / N_0}, \quad m = 1, 2, \dots, M$$

- Log-likelihood function

$$\ln f(\mathbf{y} | \mathbf{s}_m) = -\frac{N}{2} \ln(\pi N_0) - \ln \sum_{k=1}^N \frac{(y - s_{mk})^2}{N_0}, \quad m = 1, 2, \dots, M$$

- ML criterion

$$\min_m (y - s_{mk})^2, \quad m = 1, 2, \dots, M$$

- ◆ which is called *minimum (Euclidean) distance rule*

## ■ Optimum decision rule

- MAP criterion becomes ML criterion for equally probable case.
- ML criterion can be reduced to minimum Euclidean distance rule over AWGN channels.

## ■ Calculation of Euclidean distance

$$\begin{aligned} D(\mathbf{y}, \mathbf{s}_m) &= \sum_{n=1}^N y_n^2 - 2 \sum_{n=1}^N y_n s_{mn} + \sum_{n=1}^N s_{mn}^2 \\ &= \|\mathbf{y}\|^2 - 2 \mathbf{y} \cdot \mathbf{s}_m + \|\mathbf{s}_m\|^2, \quad m = 1, 2, \dots, M \end{aligned}$$

- Minimum distance rule choose  $\mathbf{s}_m$  to give the minimum distance metric which is equivalent to choose minimum value of the metric given as

$$D'(\mathbf{y}, \mathbf{s}_m) = -2 \mathbf{y} \cdot \mathbf{s}_m + \|\mathbf{s}_m\|^2, \quad m = 1, 2, \dots, M$$

or choose the maximum distance metric given as

$$C(\mathbf{y}, \mathbf{s}_m) = 2 \mathbf{y} \cdot \mathbf{s}_m - \|\mathbf{s}_m\|^2, \quad m = 1, 2, \dots, M$$



## ■ Correlation metric

$$C(\mathbf{y}, \mathbf{s}_m) = 2 \mathbf{y} \cdot \mathbf{s}_m - \|\mathbf{s}_m\|^2, \quad m = 1, 2, \dots, M$$

- We choose  $\mathbf{s}_m$  which gives maximum correlation metric.
- If all the signals have equal energy, that is,  $\|\mathbf{s}_m\|^2 = \mathcal{E}_s$ , for all  $m$ 
  - ◆ we can just neglect the term  $\|\mathbf{s}_m\|^2$ .

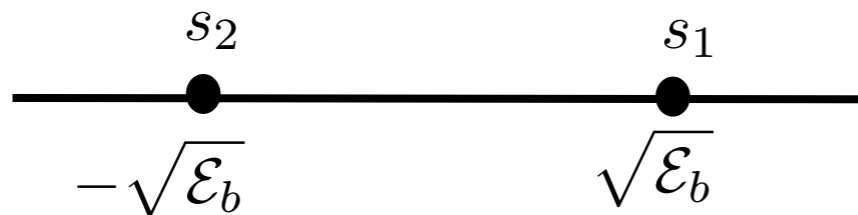
# Summary of Optimum Decision Rule

- Optimum decision rule is MAP criterion.
- MAP criterion is equivalent to ML criterion for equally probable case.
- ML criterion is equivalent to minimum Euclidean distance rule over AWGN channels.
- Minimum Euclidean distance rule is equivalent to maximum correlation rule.

# Probability of Error for M-ary Pulse Amplitude Modulation

- Bit error rate of binary PAM signals

$$P_2 = Q \left( \sqrt{\frac{2\mathcal{E}_b}{N_0}} \right)$$



$$d_{12} = 2\sqrt{\mathcal{E}_b} \implies \mathcal{E}_b = \frac{d_{12}^2}{4}$$

- We can rewrite the BER of binary PAM signals as

$$P_2 = Q \left( \sqrt{\frac{d_{12}^2}{2N_0}} \right)$$

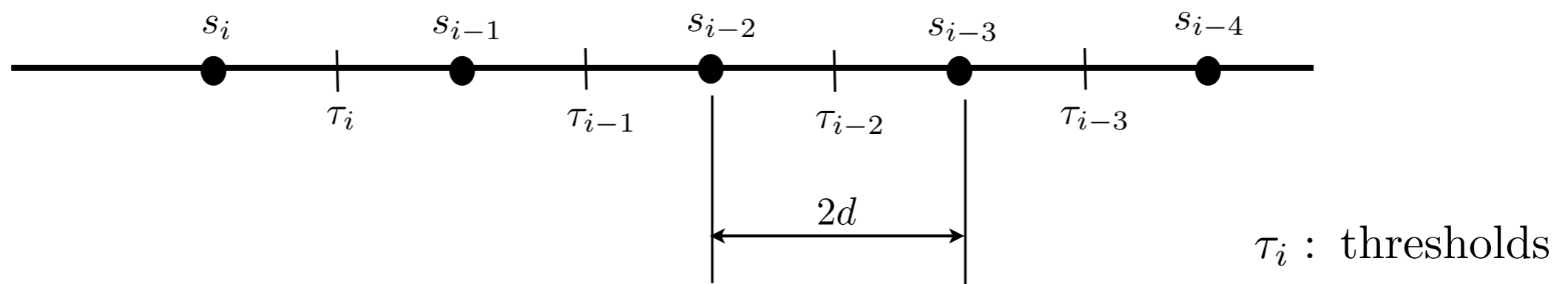
- In the case of M-ary PAM, the input to the detector is

$$y = s_m + n,$$

- Optimum decision rule for equally probable case chooses the maximum correlation metrics

$$C(y, s_m) = 2ys_m - s_m^2 = 2(y - s_m/2)s_m$$

- Equivalently, the optimum threshold may compare  $y$  with a set of M-1 thresholds, which are placed at the midpoints of successive amplitude levels. Thus, a decision is made in favor of the amplitude level that is closest to  $y$ .



■ Average symbol error rate

$$\begin{aligned} P_M &= \frac{1}{M} [\Pr(y - s_1 > d) + \Pr(|y - s_2| > d) + \cdots + \Pr(|y - s_{M-1}| > d) + \Pr(y - s_M < -d)] \\ &= \frac{M-1}{M} \Pr(|y - s_m| > d) \\ &= \frac{M-1}{M} \frac{2}{\sqrt{\pi N_0}} \int_d^\infty e^{-x^2/N_0} dx \\ &= \frac{M-1}{M} \frac{2}{\sqrt{2\pi}} \int_{\sqrt{2d^2/N_0}}^\infty e^{-x^2/2} dx \\ &= \frac{2(M-1)}{M} Q\left(\sqrt{2d^2/N_0}\right) \end{aligned}$$

● Recall

$$\mathcal{E}_{av} = \frac{d^2(M^2 - 1)}{3},$$

● We can rewrite the average symbol error rate as

$$P_M = \frac{2(M-1)}{M} Q\left(\sqrt{\frac{6\mathcal{E}_{av}}{(M^2 - 1)N_0}}\right).$$

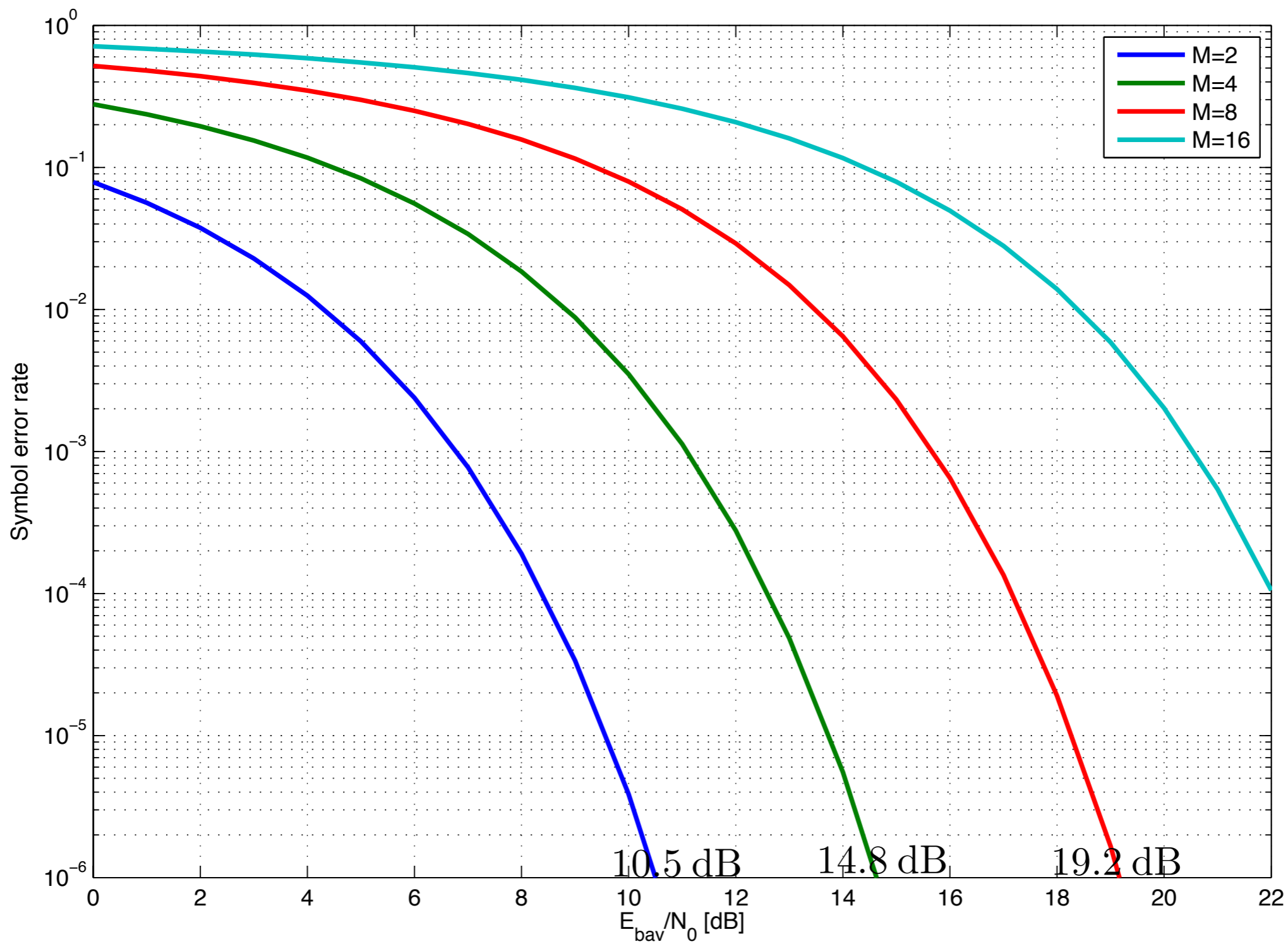
- Note that each signal carries  $k = \log_2 M$  bits of information, the average energy per bit is

$$\mathcal{E}_{bav} = \frac{\mathcal{E}_{av}}{k} \implies \mathcal{E}_{av} = k\mathcal{E}_{bav} = (\log_2 M)\mathcal{E}_{bav}$$

- Hence, we have

$$P_M = \frac{2(M-1)}{M} Q \left( \sqrt{\frac{6(\log_2 M)\mathcal{E}_{bav}}{(M^2-1)N_0}} \right).$$

■ Average SER curve



# Probability of Error for M-ary Orthogonal Signals

- Each of M-ary orthogonal signals has equal energy.
- For equal energy orthogonal signals, the optimum detector selects the signal resulting in the largest cross correlation between the received vector  $\mathbf{y}$  and each of the M possible transmitted signal vectors  $\{\mathbf{s}_m\}$ , i.e.,

$$C(\mathbf{y}, \mathbf{s}_m) = \mathbf{y} \cdot \mathbf{s}_m = \sum_{k=1}^M y_k s_{mk}, \quad m = 1, 2, \dots, M.$$

- To evaluate the probability of error, let us assume that the signal  $s_1$  is transmitted. Then the vector at the input to the detector is

$$\mathbf{y} = \left( \sqrt{E_s} + n_1, n_2, n_3, \dots, n_M \right),$$

- where  $n_1, n_2, n_3, \dots, n_M$  are zero mean, mutually statistically independent Gaussian random variables with equal variance  $N_0/2$ .



■ Cross-correlation metric

$$\begin{aligned}C(\mathbf{y}, \mathbf{s}_1) &= \sqrt{\mathcal{E}_s}(\sqrt{\mathcal{E}_s} + n_1); \\C(\mathbf{y}, \mathbf{s}_2) &= \sqrt{\mathcal{E}_s}n_2; \\&\vdots \\C(\mathbf{y}, \mathbf{s}_M) &= \sqrt{\mathcal{E}_s}n_M.\end{aligned}$$

- Note that we can eliminate the scale factor  $\sqrt{\mathcal{E}_s}$  for the comparisons.

■ PDF of the first correlator output with the elimination of  $\sqrt{\mathcal{E}_s}$ .

$$f(y_1) = \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(y_1 - \sqrt{\mathcal{E}_s})^2}{N_0}}$$

- PDF's of the other M-1 correlator outputs

$$f(y_m) = \frac{1}{\sqrt{\pi N_0}} e^{-\frac{y_m^2}{N_0}}, \quad m = 2, 3, \dots, M$$

■ Correct decision probability when  $s_1(t)$  is transmitted

$$P_{c|s_1} = \int_{-\infty}^{\infty} \Pr[n_2 < y_1, n_3 < y_1, \dots, n_M < y_1 | s_1] f_{y_1}(y_1) dy_1$$

● For equally probable case,

$$P_c = \int_{-\infty}^{\infty} \Pr(n_2 < y_1, n_3 < y_1, \dots, n_M < y_1 | y_1) f_{y_1}(y_1) dy_1$$

● Note that

$$\begin{aligned} \Pr(n_M < y_1 | y_1) &= \int_{-\infty}^{y_1} f(y_m) dy_m, \quad m = 2, 3, \dots, M \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\frac{\sqrt{2y_1^2}}{N_0}} e^{-\frac{y_m^2}{2}} dy_m \\ &= 1 - Q\left(\sqrt{\frac{2y_1^2}{N_0}}\right). \end{aligned}$$

- Probability of correct decision

$$P_c = \int_{-\infty}^{\infty} \left[ 1 - Q \left( \sqrt{\frac{2y_1^2}{N_0}} \right) \right]^{M-1} f_{y_1}(y_1) dy_1$$

- Probability of symbol error (Symbol error rate)

$$P_M = 1 - \int_{-\infty}^{\infty} \left[ 1 - Q \left( \sqrt{\frac{2y_1^2}{N_0}} \right) \right]^{M-1} \frac{1}{\sqrt{\pi N_0}} e^{-(y-\sqrt{\mathcal{E}_s})^2/N_0} dy_1$$

Change of variable  $x = \frac{2y_1^2}{N_0}$

Then we have

$$\begin{aligned} P_M &= 1 - \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} [1 - Q(x)]^{M-1} e^{-(x-\sqrt{2\mathcal{E}_s/N_0})} dx \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left\{ 1 - [1 - Q(x)]^{M-1} \right\} e^{-(x-\sqrt{2\mathcal{E}_s/N_0})} dx \end{aligned}$$

■ Average bit error rate

$$P_b = \frac{1}{k} \sum_{n=1}^k n \binom{n}{k} \frac{P_M}{2^k - 1} = \frac{2^{k-1}}{2^k - 1} P_M \approx \frac{P_M}{2}, \quad k \gg 1.$$