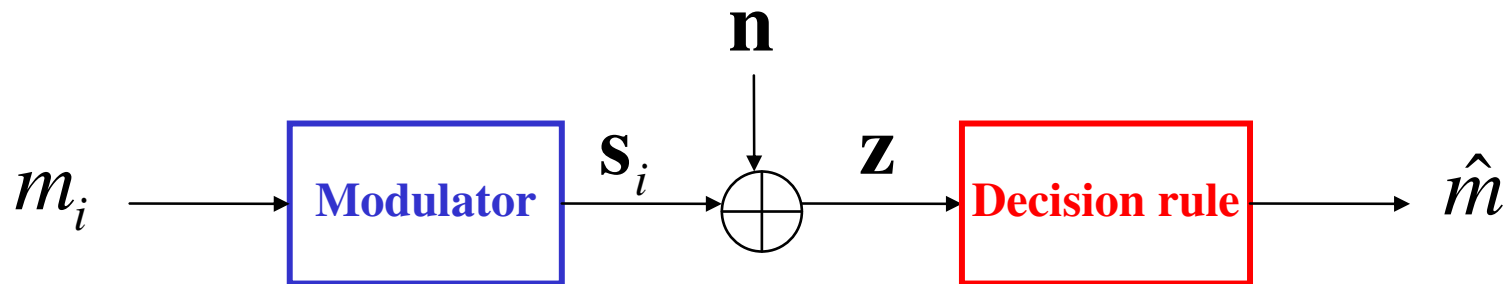


Detection of signal in AWGN

- Detection problem:
 - Given the observation vector \mathbf{z} , perform a mapping from \mathbf{z} to an estimate \hat{m} of the transmitted symbol, m_i , such that the average probability of error in the decision is minimized.



Statistics of the observation Vector

- AWGN channel model: $\mathbf{z} = \mathbf{s}_i + \mathbf{n}$
 - Signal vector $\mathbf{s}_i = (a_{i1}, a_{i2}, \dots, a_{iN})$ is deterministic.
 - Elements of noise vector $\mathbf{n} = (n_1, n_2, \dots, n_N)$ are i.i.d Gaussian random variables with zero-mean and variance $N_0 / 2$. The noise vector pdf is

$$p_{\mathbf{n}}(\mathbf{n}) = \frac{1}{(\pi N_0)^{N/2}} \exp\left(-\frac{\|\mathbf{n}\|^2}{N_0}\right)$$

- The elements of observed vector $\mathbf{z} = (z_1, z_2, \dots, z_N)$ are independent Gaussian random variables. Its pdf is

$$p_{\mathbf{z}}(\mathbf{z} | \mathbf{s}_i) = \frac{1}{(\pi N_0)^{N/2}} \exp\left(-\frac{\|\mathbf{z} - \mathbf{s}_i\|^2}{N_0}\right)$$

Detection

- Optimum decision rule (maximum a posteriori probability):

Set $\hat{m} = m_i$ if

$\Pr(m_i \text{ sent} | \mathbf{z}) \geq \Pr(m_k \text{ sent} | \mathbf{z})$, for all $k \neq i$

where $k = 1, \dots, M$.

- Applying Bayes' rule gives:

Set $\hat{m} = m_i$ if

$p_k \frac{p_{\mathbf{z}}(\mathbf{z} | m_k)}{p_{\mathbf{z}}(\mathbf{z})}$, is maximum for all $k = i$

Detection ...

- Partition the signal space into M decision regions, Z_1, \dots, Z_M such that

Vector \mathbf{z} lies inside region Z_i if

$$\ln\left[p_k \frac{p_{\mathbf{z}}(\mathbf{z} | m_k)}{p_{\mathbf{z}}(\mathbf{z})}\right], \text{ is maximum for all } k = i.$$

That means

$$\hat{m} = m_i$$

Detection (ML rule)

- For equal probable symbols, the optimum decision rule (maximum posteriori probability) is simplified to:

Set $\hat{m} = m_i$ if
 $p_{\mathbf{z}}(\mathbf{z} | m_k)$, is maximum for all $k = i$

or equivalently:

Set $\hat{m} = m_i$ if
 $\ln[p_{\mathbf{z}}(\mathbf{z} | m_k)]$, is maximum for all $k = i$

which is known as *maximum likelihood*.

Detection (ML)...

- Partition the signal space into M decision regions, Z_1, \dots, Z_M .
- Restate the maximum likelihood decision rule as follows:

Vector \mathbf{z} lies inside region Z_i if
 $\ln[p_{\mathbf{z}}(\mathbf{z} | m_k)]$, is maximum for all $k = i$
That means
 $\hat{m} = m_i$

Detection rule (ML)...

- It can be simplified to:

Vector \mathbf{z} lies inside region Z_i if $\|\mathbf{z} - \mathbf{s}_k\|$, is minimum for all $k = i$

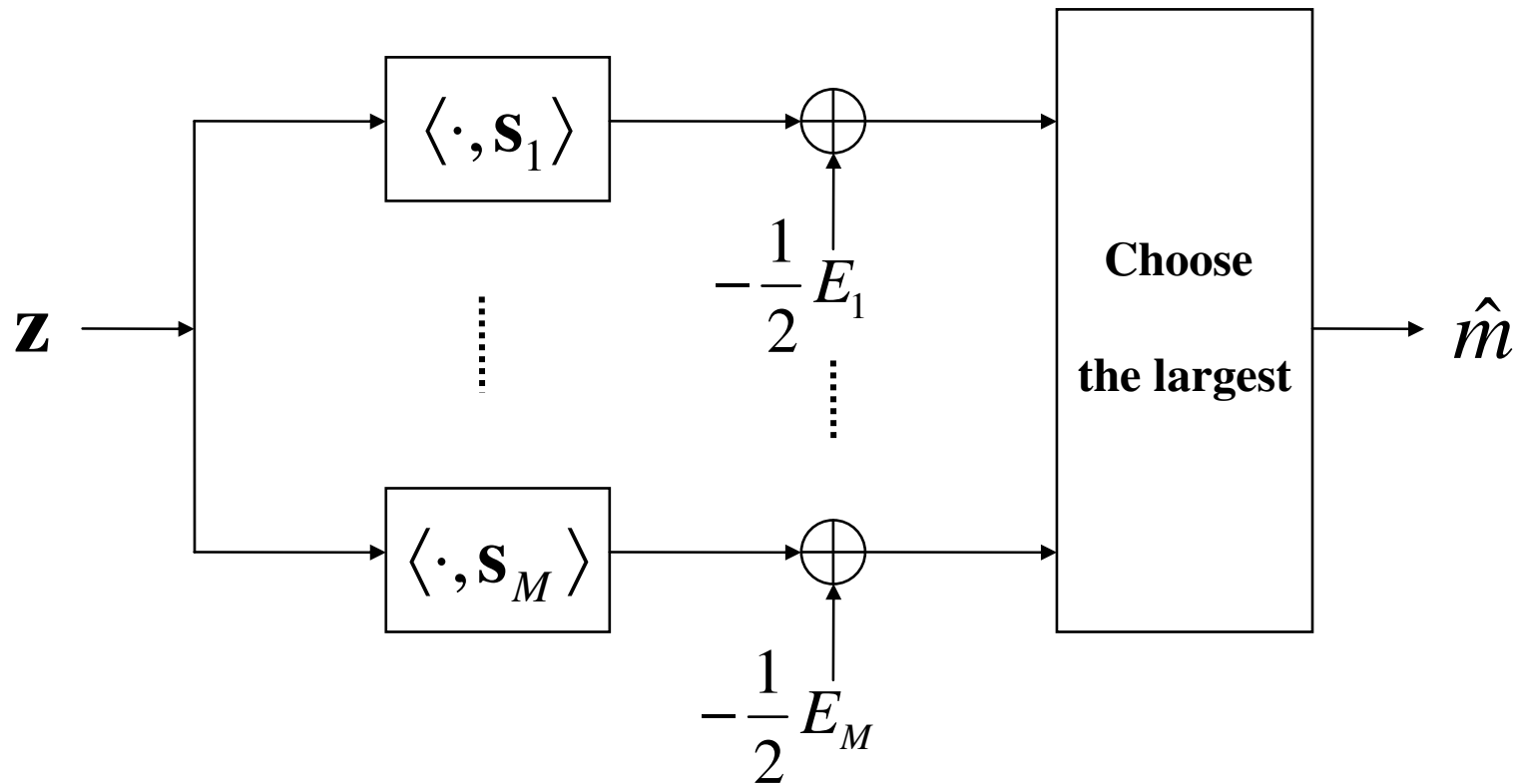
or equivalently:

Vector \mathbf{r} lies inside region Z_i if

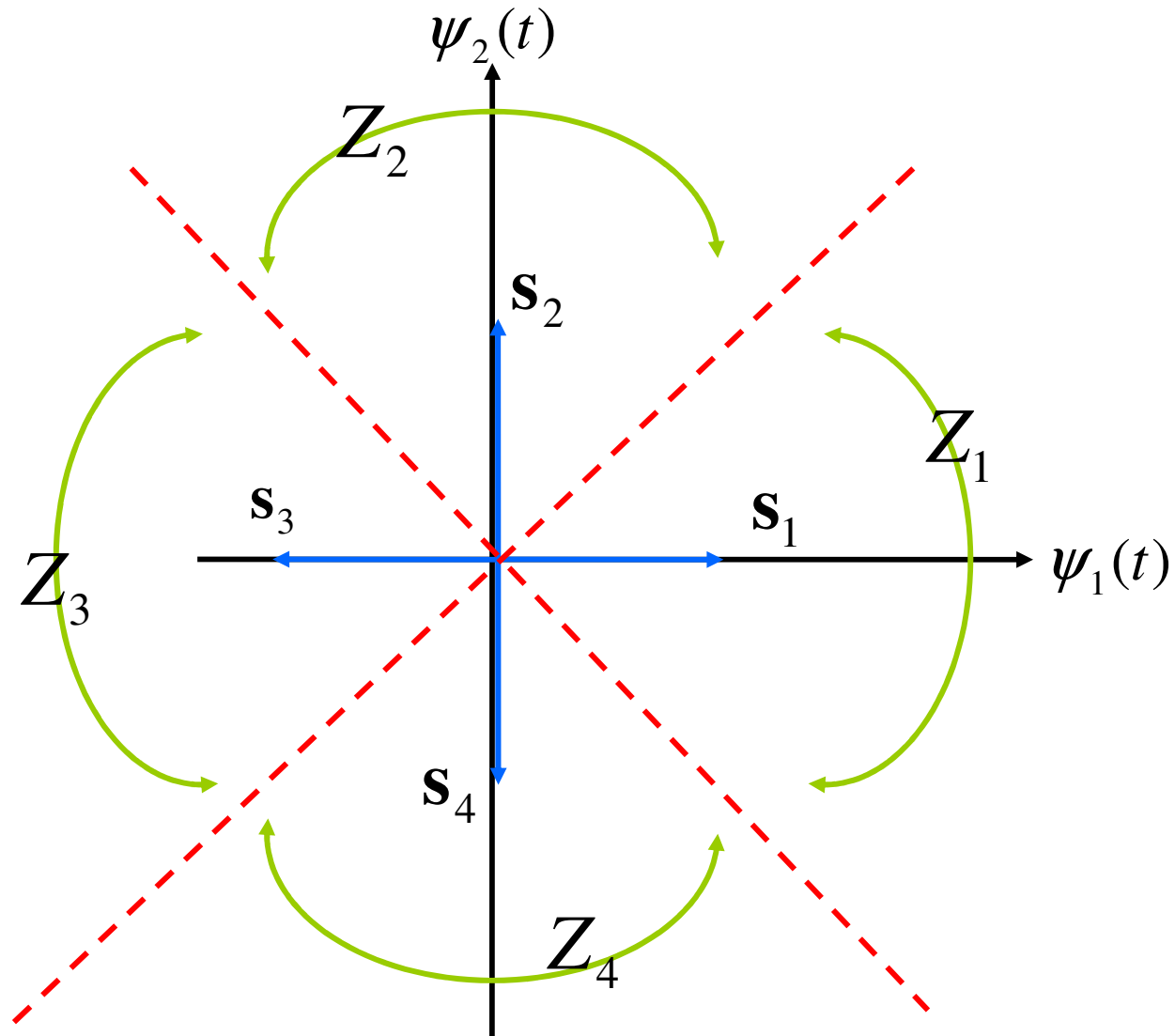
$$\sum_{j=1}^N z_j a_{kj} - \frac{1}{2} E_k, \text{ is maximum for all } k = i$$

where E_k is the energy of $s_k(t)$.

Maximum likelihood detector block diagram



Schematic example of ML decision regions



Average probability of symbol error

- **Erroneous decision:** For the transmitted symbol m_i or equivalently signal vector \mathbf{S}_i , an error in decision occurs if the observation vector \mathbf{z} does not fall inside region Z_i .

- Probability of erroneous decision for a transmitted symbol

$$P_e(m_i) = \Pr(\hat{m} \neq m_i \text{ and } m_i \text{ sent})$$

or equivalently

$$\Pr(\hat{m} \neq m_i) = \Pr(m_i \text{ sent})\Pr(\mathbf{z} \text{ does not lie inside } Z_i | m_i \text{ sent})$$

- Probability of correct decision for a transmitted symbol

$$\Pr(\hat{m} = m_i) = \Pr(m_i \text{ sent})\Pr(\mathbf{z} \text{ lies inside } Z_i | m_i \text{ sent})$$

$$P_c(m_i) = \Pr(\mathbf{z} \text{ lies inside } Z_i | m_i \text{ sent}) = \int_{Z_i} p_{\mathbf{z}}(\mathbf{z} | m_i) d\mathbf{z}$$

$$P_e(m_i) = 1 - P_c(m_i)$$

Av. prob. of symbol error ...

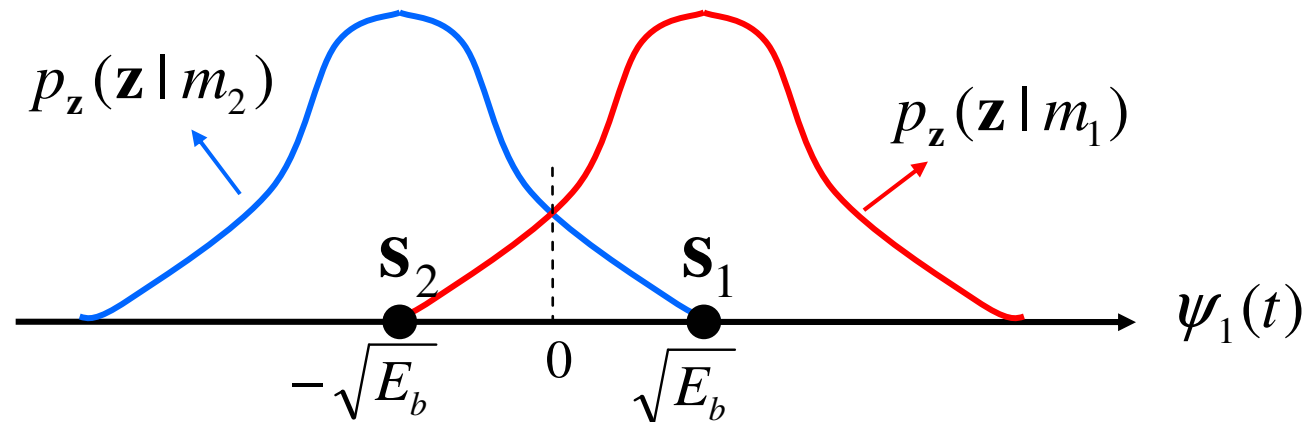
- Average probability of symbol error :

$$P_E(M) = \sum_{i=1}^M \Pr(\hat{m} \neq m_i)$$

- For equally probable symbols:

$$\begin{aligned} P_E(M) &= \frac{1}{M} \sum_{i=1}^M P_e(m_i) = 1 - \frac{1}{M} \sum_{i=1}^M P_c(m_i) \\ &= 1 - \frac{1}{M} \sum_{i=1}^M \int_{Z_i} p_{\mathbf{z}}(\mathbf{z} | m_i) d\mathbf{z} \end{aligned}$$

Example for binary PAM



$$P_e(m_1) = P_e(m_2) = Q\left(\frac{\|\mathbf{s}_1 - \mathbf{s}_2\|/2}{\sqrt{N_0/2}}\right)$$

$$P_B = P_E(2) = Q\left(\sqrt{\frac{2E_b}{N_0}}\right)$$

Union bound

Union bound

The probability of a finite union of events is upper bounded by the sum of the probabilities of the individual events.

- Let A_{ki} denote that the observation vector \mathbf{Z} is closer to the symbol vector \mathbf{s}_k than \mathbf{s}_i , when \mathbf{s}_i is transmitted.
- $\Pr(A_{ki}) = P_2(\mathbf{s}_k, \mathbf{s}_i)$ depends only on \mathbf{s}_i and \mathbf{s}_k .
- Applying Union bounds yields

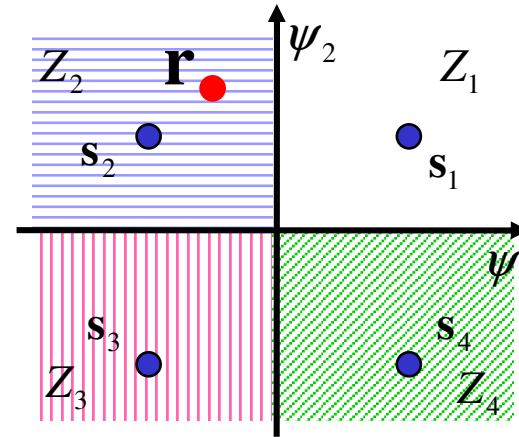
$$P_e(m_i) \leq \sum_{\substack{k=1 \\ k \neq i}}^M P_2(\mathbf{s}_k, \mathbf{s}_i)$$



$$P_E(M) \leq \frac{1}{M} \sum_{i=1}^M \sum_{\substack{k=1 \\ k \neq i}}^M P_2(\mathbf{s}_k, \mathbf{s}_i)$$

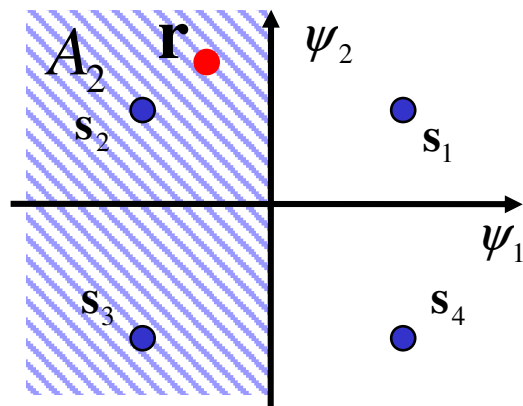
Example of union bound

$$P_e(m_1) = \int_{Z_2 \cup Z_3 \cup Z_4} p_{\mathbf{r}}(\mathbf{r} | m_1) d\mathbf{r}$$

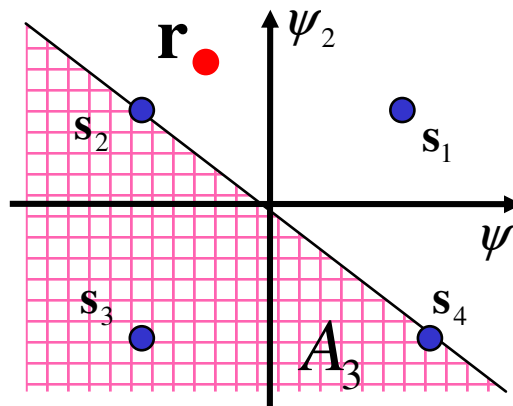


Union bound:

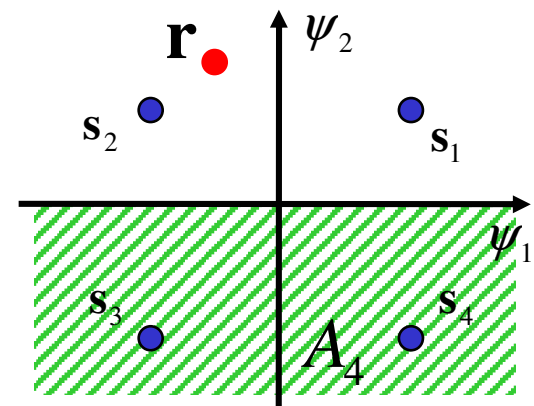
$$P_e(m_1) \leq \sum_{k=2}^4 P_2(\mathbf{s}_k, \mathbf{s}_1)$$



$$P_2(\mathbf{s}_2, \mathbf{s}_1) = \int_{A_2} p_{\mathbf{r}}(\mathbf{r} | m_1) d\mathbf{r}$$



$$P_2(\mathbf{s}_3, \mathbf{s}_1) = \int_{A_3} p_{\mathbf{r}}(\mathbf{r} | m_1) d\mathbf{r}$$



$$P_2(\mathbf{s}_4, \mathbf{s}_1) = \int_{A_4} p_{\mathbf{r}}(\mathbf{r} | m_1) d\mathbf{r}$$

Upper bound based on minimum distance

$P_2(\mathbf{s}_k, \mathbf{s}_i) = \Pr(\mathbf{z} \text{ is closer to } \mathbf{s}_k \text{ than } \mathbf{s}_i, \text{ when } \mathbf{s}_i \text{ is sent})$

$$= \int_{d_{ik}}^{\infty} \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{u^2}{N_0}\right) du = Q\left(\frac{d_{ik}/2}{\sqrt{N_0}/2}\right)$$

$$d_{ik} = \|\mathbf{s}_i - \mathbf{s}_k\|$$

$$P_E(M) \leq \frac{1}{M} \sum_{i=1}^M \sum_{\substack{k=1 \\ k \neq i}}^M P_2(\mathbf{s}_k, \mathbf{s}_i) \leq (M-1) Q\left(\frac{d_{\min}/2}{\sqrt{N_0}/2}\right)$$

Minimum distance in the signal space: $d_{\min} = \min_{\substack{i,k \\ i \neq k}} d_{ik}$

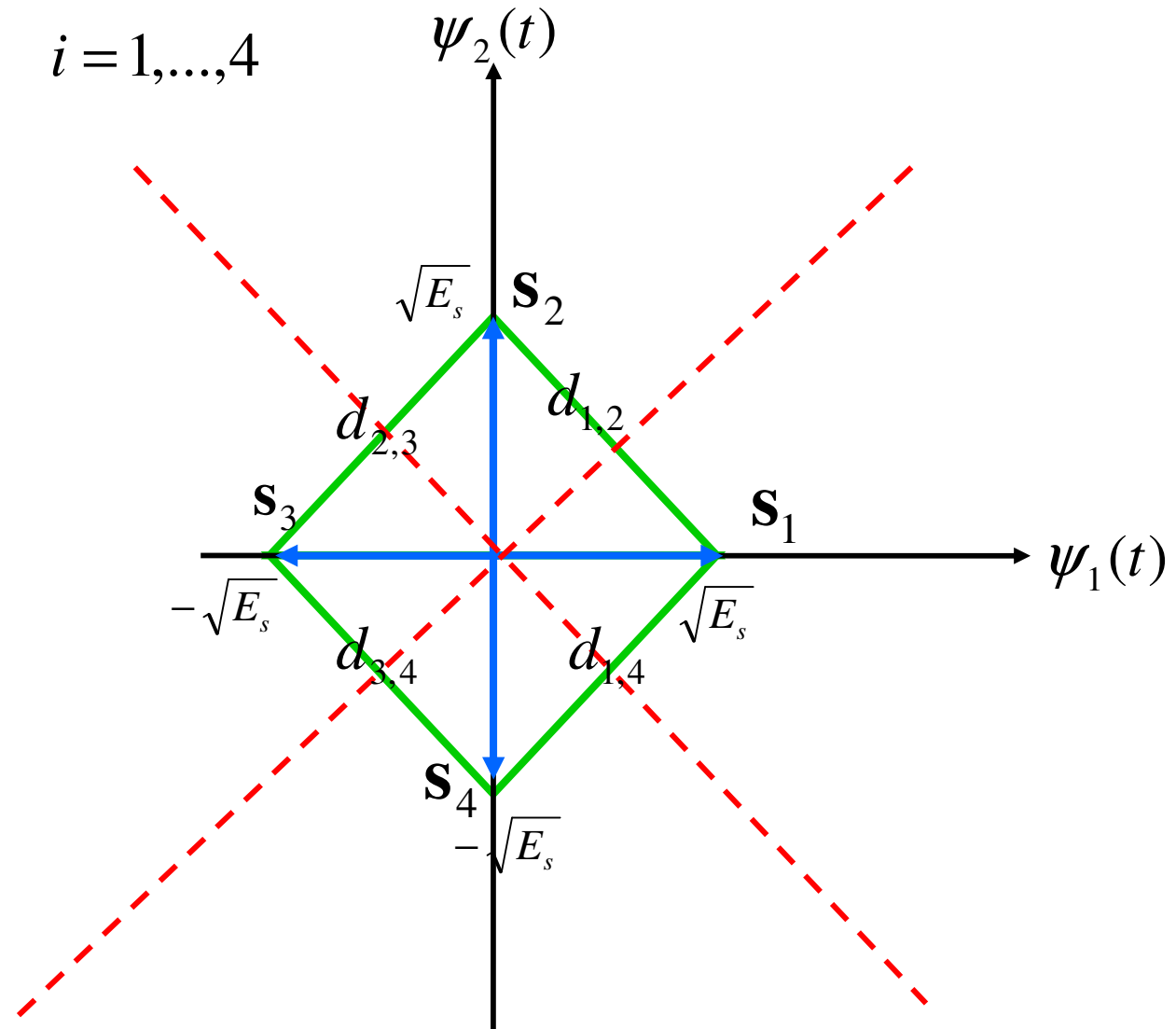
Example of upper bound on av. Symbol error prob. based on union bound

$$\|\mathbf{s}_i\| = \sqrt{E_i} = \sqrt{E_s}, \quad i = 1, \dots, 4$$

$$d_{i,k} = \sqrt{2E_s}$$

$$i \neq k$$

$$d_{\min} = \sqrt{2E_s}$$



Eb/No figure of merit in digital communications

- SNR or S/N is the average signal power to the average noise power. SNR should be modified in terms of bit-energy in DCS, because:
 - Signals are transmitted within a symbol duration and hence, are energy signal (zero power).
 - A merit at bit-level facilitates comparison of different DCSs transmitting different number of bits per symbol.

$$\frac{E_b}{N_0} = \frac{S T_b}{N / W} = \frac{S}{N} \frac{W}{R_b}$$

R_b : Bit rate

W : Bandwidth

Example of Symbol error prob. For PAM signals

