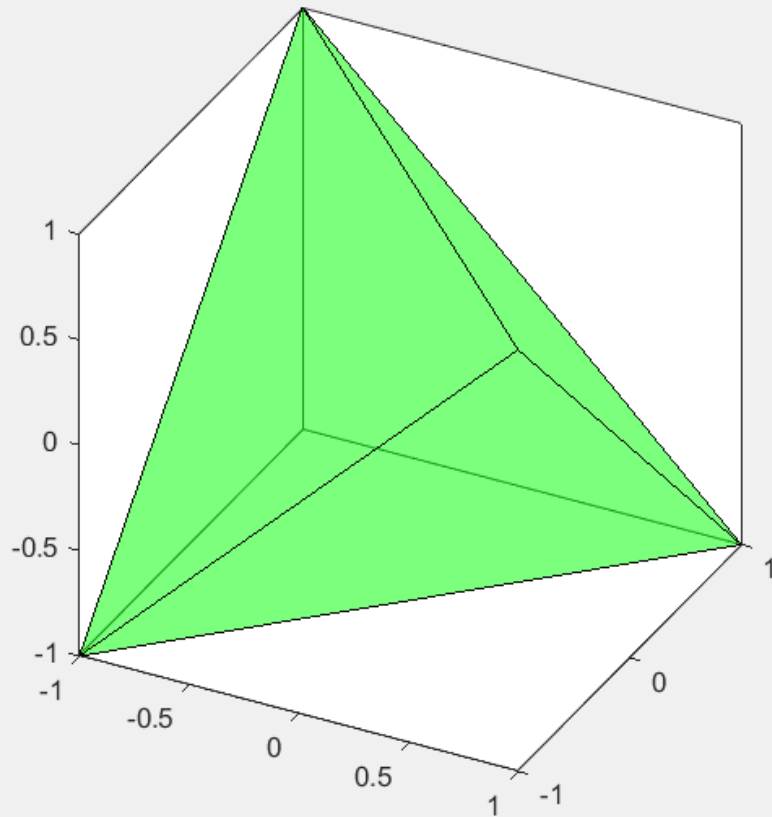
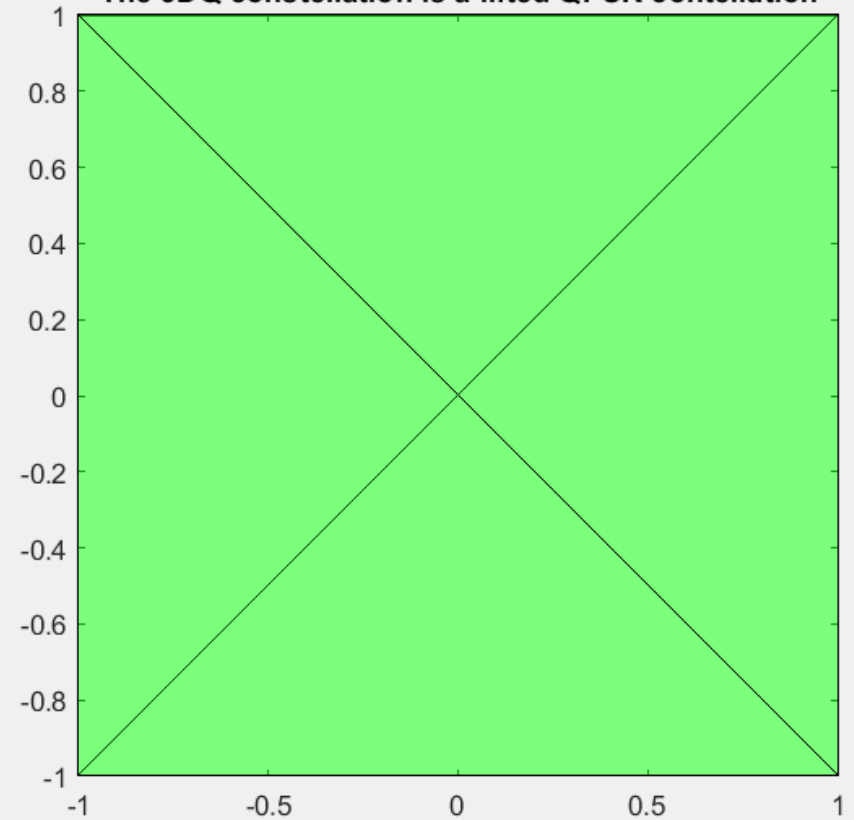


The 3DQ constellation: Vertices of a tetrahedron!



The 3DQ constellation is a lifted QPSK constellation



MATLAB script: "plot\_3DQ\_constellation.m"

Recall that the syndrome vector, computed from the received vector as  $\bar{s} = \bar{r}H^T$  is related to the error vector by the equation

$$\bar{s} = \bar{e}H^T, \quad (1)$$

where  $H$  is the parity-check matrix. Note that syndrome  $\bar{s}$  is an  $(n-k)$ -bit vector.

Eq. (1) above indicates that the syndrome is a function of the error vector, i.e., there is a one-to-one relation between the error vector  $\bar{e}$  and the syndrome  $\bar{s}$ .

## The Hamming bound

$$\bar{s} = \bar{e}H^T$$

“To correct up to  $t$  errors, the number of different syndrome values  $\bar{s}$  needs to be greater than or equal to the number of different error vectors  $\bar{e}$  with up to  $t$  ones.”

This gives the *Hamming bound*:

$$2^{n-k} \geq \sum_{i=0}^t \binom{n}{i}$$

For  $t = 1$ , or *single-error correcting codes*, the minimum number of different syndrome values is obtained as

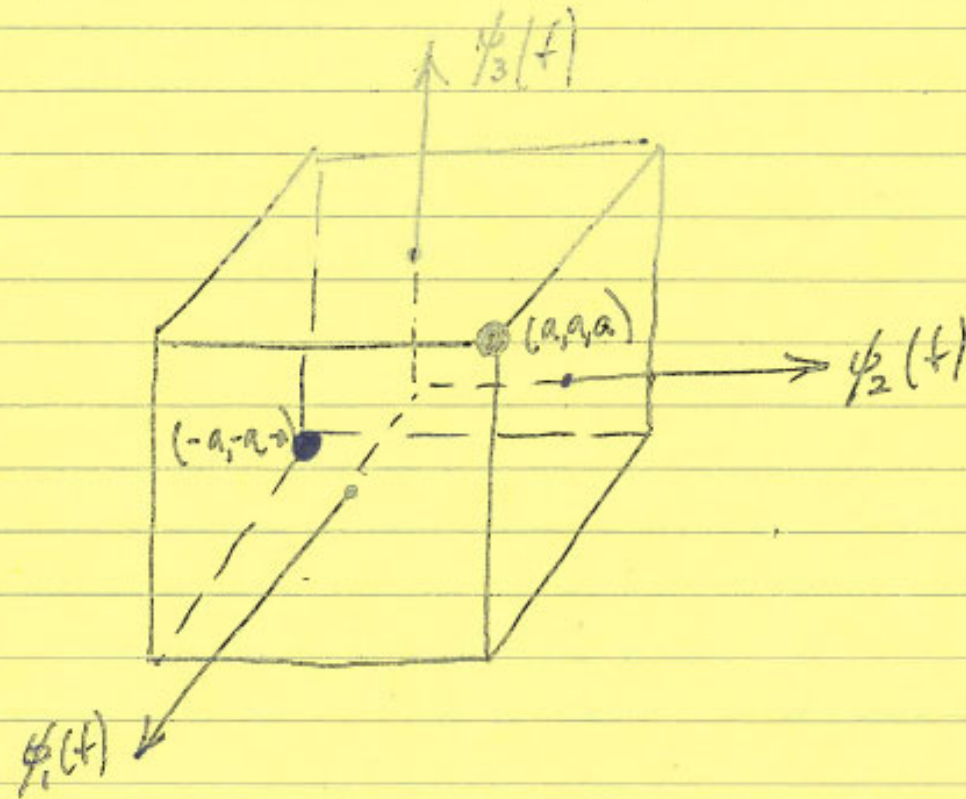
$$2^{n-k} = \sum_{i=0}^1 \binom{n}{i} = 1 + n \quad (2)$$

For a single-bit error correcting code, an equivalent statement is that **all the columns of  $H$  be different**. This idea due to Hamming (1949) gives binary linear *Hamming codes*. With  $m=n-k$  equal to the number of redundant bits, from Eq. (2) *Hamming codes* have parameters:

Length:	$n = 2^m - 1$
Dimension:	$k = 2^m - 1 - m$
Minimum Hamming distance:	$d_{\min} = 3$

**The smallest Hamming code is obtained with  $m=2$  and gives the binary linear (3,1,3) repetition code!**

Another example:  $(3,1,3)$  Binary repetition code



Example: Binary repetition (3,1,3) code

$$G = (111), \quad H = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}$$

<u>Error patterns, <math>\bar{e}</math></u>	<u>Syndrome, <math>\bar{s} = \bar{e}H^T</math></u>
(000)	(00)
(100)	(11)
(010)	(10)
(001)	(01)

Example: If  $\bar{C} = (1\ 1\ 1)$  and  $F = (1\ 1\ 0)$   
 then  $\bar{S} = \bar{F} H^T$

$$= (1\ 1\ 0) \begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 1 \end{pmatrix} = (0\ 1)$$

$\longleftrightarrow \bar{e} = (0\ 0\ 1)$

From table

$$\therefore \hat{C} = \bar{r} \oplus \bar{e} = \begin{pmatrix} 1 & 1 & 0 \\ \oplus & (0 & 0 & 1) \end{pmatrix} = (1\ 1\ 1)$$



Hamming (7,4,3) code:

All nonzero column vectors of 3 bits

$$H = \begin{bmatrix} 1 & 0 & 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 & 1 \end{bmatrix} = [P^T \mid I_3]$$

Therefore

$$G = [I_4 \mid P] = \begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix}$$

Suppose information bits  $\bar{B} = (0110)$ . Then

$$\bar{C} = (0110) \begin{bmatrix} G \\ + \end{bmatrix} = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 \end{pmatrix} = \underline{\underline{(0110110)}}$$

Suppose error in 4th position, so that

$$\bar{r} = (0 \ 1 \ 1 \ \underset{\substack{\uparrow \\ \text{error}}}{1} \ 1 \ 1 \ 0)$$

① Compute syndrome:

$$\boxed{\bar{s} = \bar{r} H^T} = (0 \ 1 \ 1 \ 1 \ 1 \ 0) \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \underline{\underline{(1 \ 1 \ 1)}}.$$

Homework:  
Write as equations/circuit

② This corresponds to the 4th column of  $H$ .  
 $\uparrow$  transpose of the  
 $\therefore$  Error in 4th position

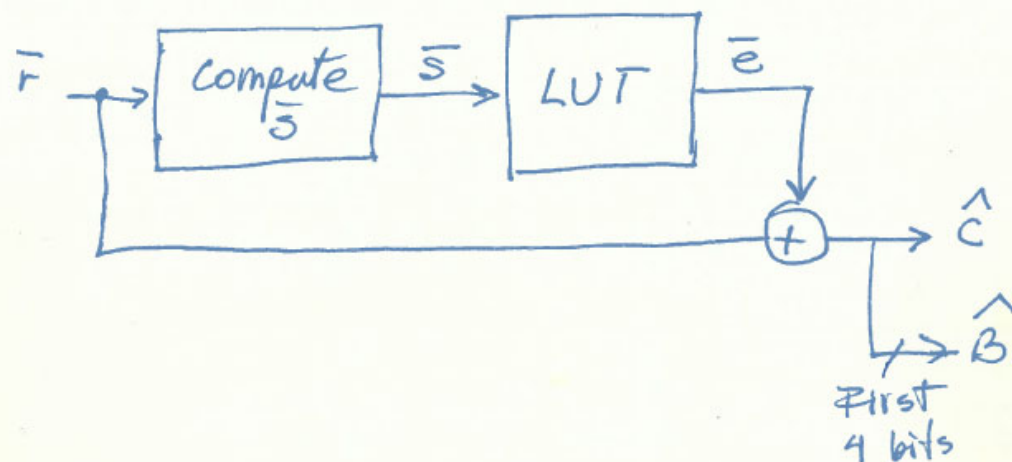
$$\hat{c} = (0 \ 1 \ 1 \ 0 \ 1 \ 1 \ 0) \rightarrow (\underbrace{0 \ 1 \ 1}_\text{Information bits}) = \hat{B}$$

Look-up Table :

Syndrome $\bar{s}$	Error vector $\bar{e}$
000	00000000
110	10000000
011	01000000
101	00100000
111	00010000
100	00001000
010	00000100
001	00000010

$H^T$  →

H-D decoder :



# Coded BPSK communication

- Codeword  $\bar{c} = (c_1 \ c_2 \ \cdots \ c_n)$  from a binary  $(n, k, d_{\min})$  code is transmitted with BPSK modulation (polar mapping) as a vector of **mapper** outputs (signal points)  $\bar{s} = (s_1 \ s_2 \ \cdots \ s_n)$ , where  $s_i = (2c_i - 1)a$ , for  $i = 1, 2, \dots, n$  and  $a = \sqrt{E_s/n}$
- Zero-mean **noise**  $N_i$ , with  $\sigma_i^2 = N_0/2$ ,  $i = 1, 2, \dots, n$ , added at the receiver
- **Matched filter** outputs are  $Y_i = s_i + N_i$ ,  $i = 1, 2, \dots, n$
- **Likelihood:**

$$L(\bar{c}) = p(\bar{Y}|\bar{s}) = \frac{1}{(\sqrt{\pi N_0})^n} \exp \left( -\frac{1}{N_0} \sum_{i=1}^n (Y_i - s_i)^2 \right),$$

which is **maximized** by the simplified log likelihood or **correlation**:

$$\ell(\bar{c}) = \sum_{i=1}^n s_i Y_i$$

The log-likelihood (log of the conditional prob.)  $\log(p(\bar{Y}|\bar{s})) = L(\bar{c})$  (\*)

can be written as

$$L(\bar{c}) = \sum_{i=1}^n s_i Y_i$$

$$\boxed{L(\bar{c}) = \sum_{i=1}^n m(c_i) Y_i}, \quad \underline{\text{correlation}}$$

where  $Y_i$  denote MF outputs

Soft-decision decoding

Choose as the most likely transmitted codeword  $\bar{c} = \bar{B}G$ , the one that maximizes  $L(\bar{c})$

$$(*) \quad p(\bar{Y}|\bar{s}) = \frac{1}{(\sqrt{\pi N_0})^n} \cdot e^{-\frac{1}{N_0} \sum_{i=1}^n (Y_i - s_i)^2}$$

$s_i = \pm\sqrt{E}$  for BPSK (polar)

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CORRELATIONS OF (3,2,2)  
EPC CODE

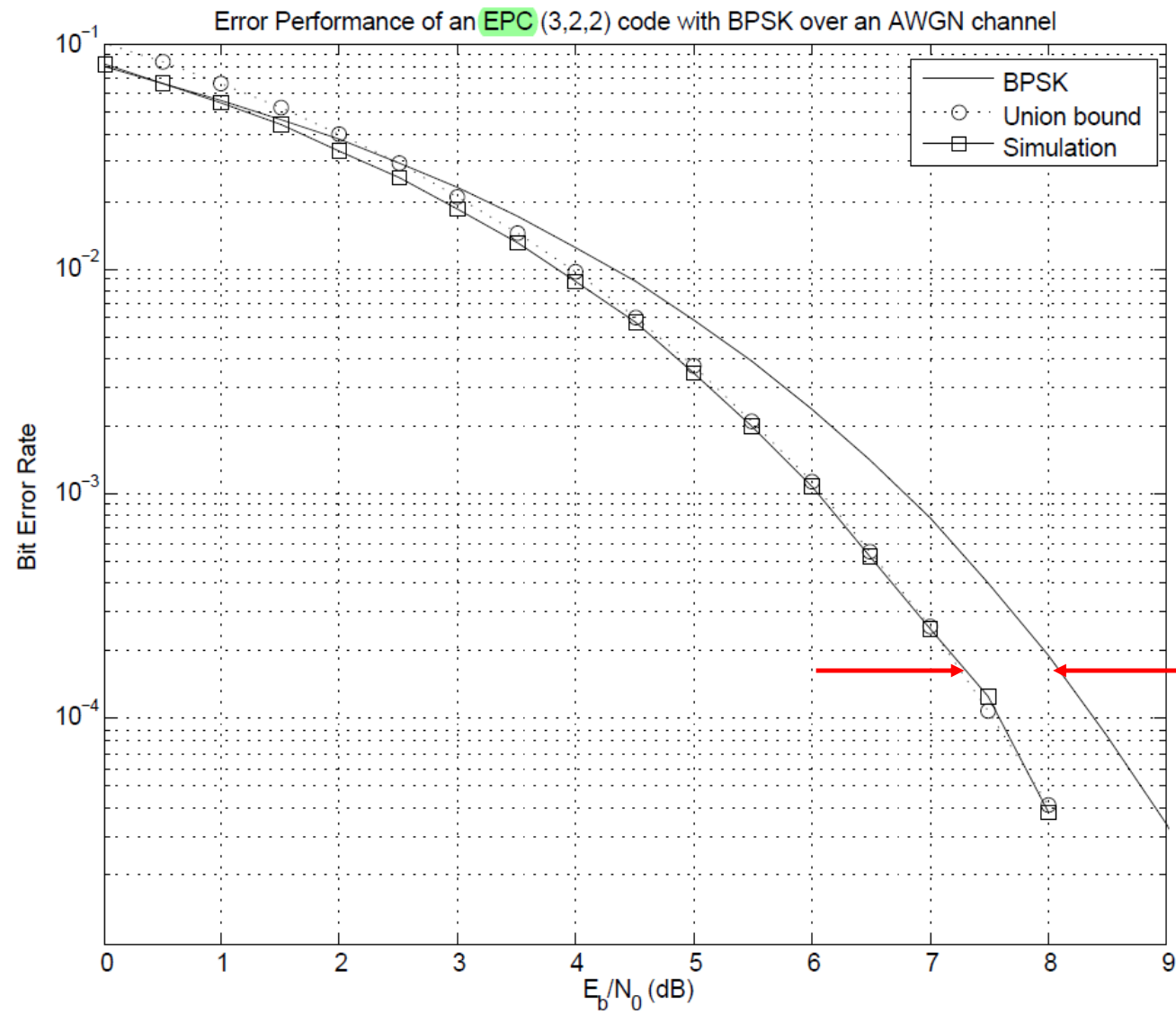
Uncoded $B_1 B_2$	Coded $C_1 C_2 C_3$	$\bar{S}$	Correlations $P$	Map: $0 \rightarrow +1$ $1 \rightarrow -1$
0 0	0 0 0	a a a	$Y_1 + Y_2 + Y_3$	
0 1	0 1 1	a -a -a	$Y_1 - Y_2 - Y_3$	
1 1	1 1 0	-a -a a	$-Y_1 - Y_2 + Y_3$	
1 0	1 0 1	-a a -a	$-Y_1 + Y_2 - Y_3$	

$$a = \sqrt{E_s/3}$$

MAP:  $0 \rightarrow -a$  $1 \rightarrow +a$ 

$B_1 B_2$	$C_1 C_2 C_3$	$\bar{S}$	Correlations $P$
0 0	0 0 0	-a -a -a	$-Y_1 - Y_2 - Y_3$
0 1	0 1 1	-a +a +a	$-Y_1 + Y_2 + Y_3$
1 1	1 1 0	+a +a -a	$+Y_1 + Y_2 - Y_3$
1 0	1 0 1	+a -a +a	$+Y_1 - Y_2 + Y_3$





$\approx 0.93$  dB

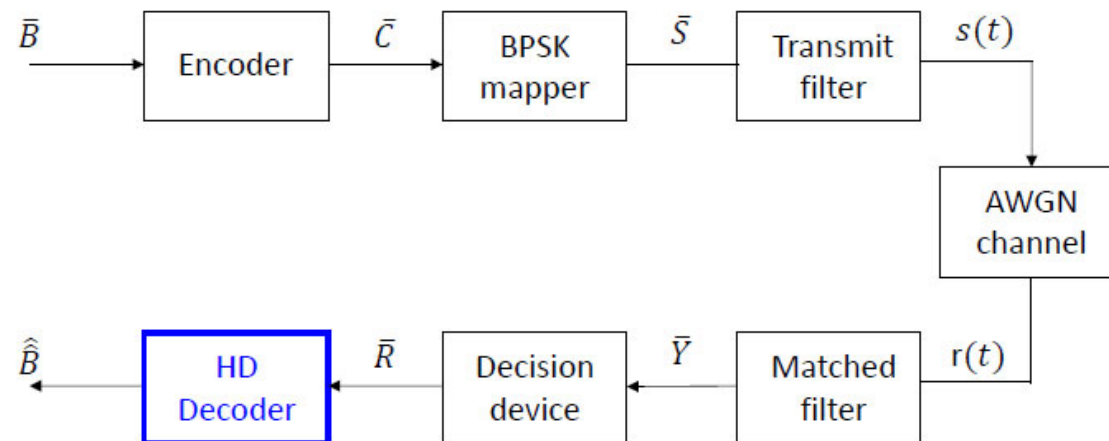
## Hard-decision (HD) decoding versus soft-decision (SD) decoding

- HD decoding: Correct errors after ML estimation (decision device),  $R_k$
- SD decoding: Use of matched-filter outputs,  $Y_k$

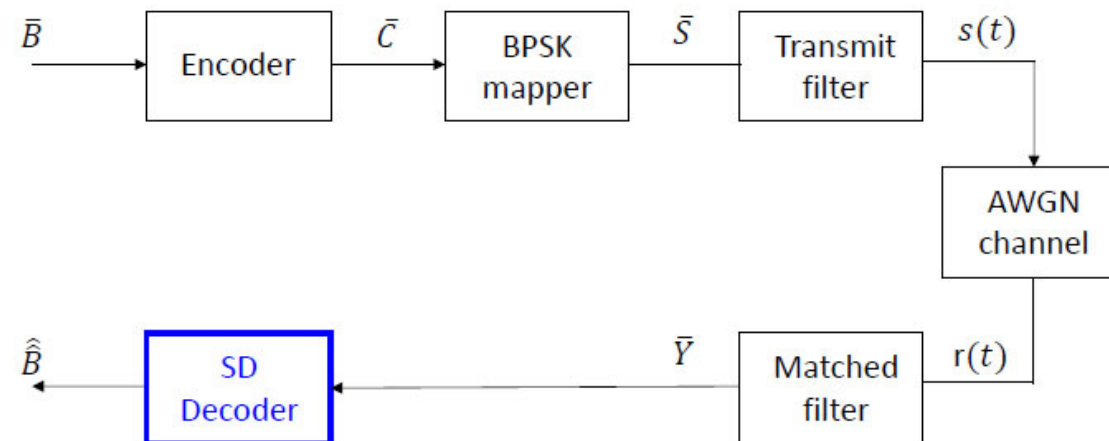
**SD decoding performs better than HD decoding**, as the *combined likelihood of all coded bits* is used instead of bit-by-bit decisions



## Hard-decision decoding

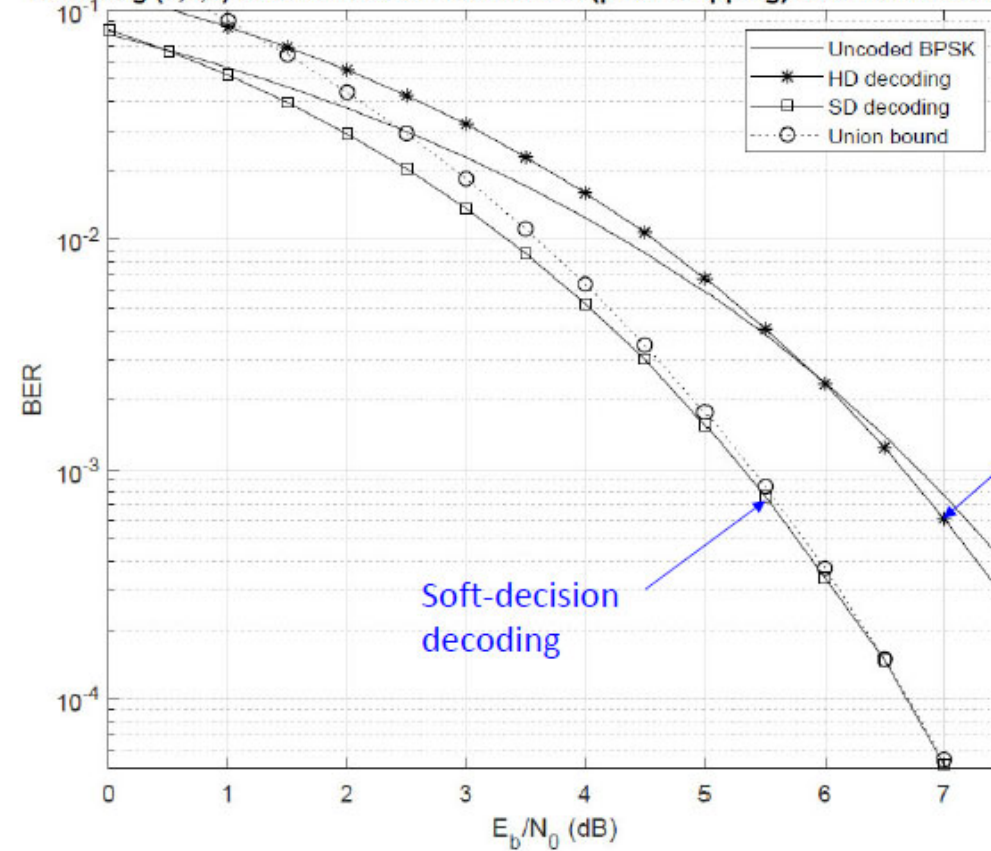


## Soft-decision decoding



# Example: Binary Hamming (7,4,3) code

Hamming (7,4,3) code with BPSK modulation (polar mapping) over an AWGN channel



Hard-decision  
decoding

Soft-decision  
decoding

## Coding gain and "real" (effective) coding gain

Binary  $(n, k, d_{\min})$  codes with polar mapping over AWGN

Modulated codeword :  $\bar{s} = (\pm a, \pm a, \dots, \pm a)$

$$= a \cdot ((-1)^{b_1}, (-1)^{b_2}, \dots, (-1)^{b_n})$$

Average symbol energy :  $n \cdot a^2 = E_s \Rightarrow a = \sqrt{E_s/n}$

Each codeword carries  $k$  bits. therefore  $E_s = k E_b$   
or

$$a = \sqrt{\frac{k}{n} E_b}$$

note, this is  
the pulse energy

The <sup>min. Euclidean</sup> distance between  $k$  codewords modulated is  $d_{\min}^E = \sqrt{4d_{\min} \cdot a^2}$   
 $\sqrt{d_{\min} (2a)^2}$

Finally, the pairwise error probability between modulated codewords is

$$P_2 = Q \left( \sqrt{d_{\min} \frac{k}{n} \frac{2E_b}{N_0}} \right) = Q \left( \sqrt{\frac{d^2}{2N_0}} \right)$$

At high values of  $E_b/N_0$ , the average probability of a bit error

$$P_b \approx Q \left( \sqrt{d_{\min} \cdot \frac{k}{n} \cdot \frac{2E_b}{N_0}} \right) = Q \left( \sqrt{\frac{R \cdot d}{n \cdot N_0}} \right)$$

ASYMPTOTIC CODING GAIN = CG

Code	$d_{\min}$	$k/n$	CG (dB)
(3, 2, 2)	2	2/3	1.25
(4, 3, 2)	2	3/4	1.76
(7, 4, 3)	3	4/7	2.34



This asymptotic coding gain does not take into account the number of codewords at minimum distance,  $A_{\min}$ .

The loss due to this number is approx.  $0.2 \log_2(A_{\min})$  so that the coding gain can be better as

Real  
Coding  
Gain

$$RCG \approx 10 \log_{10} \left( d_{\min} \cdot \frac{k}{n} \right) - 0.2 \log_2(A_{\min}) \quad (\text{dB})$$

Code	RCG	$A_{\min}$
$(3,2,2)$	0.93	3
$(4,3,2)$	1.24	6
$(7,4,3)$	1.78	7