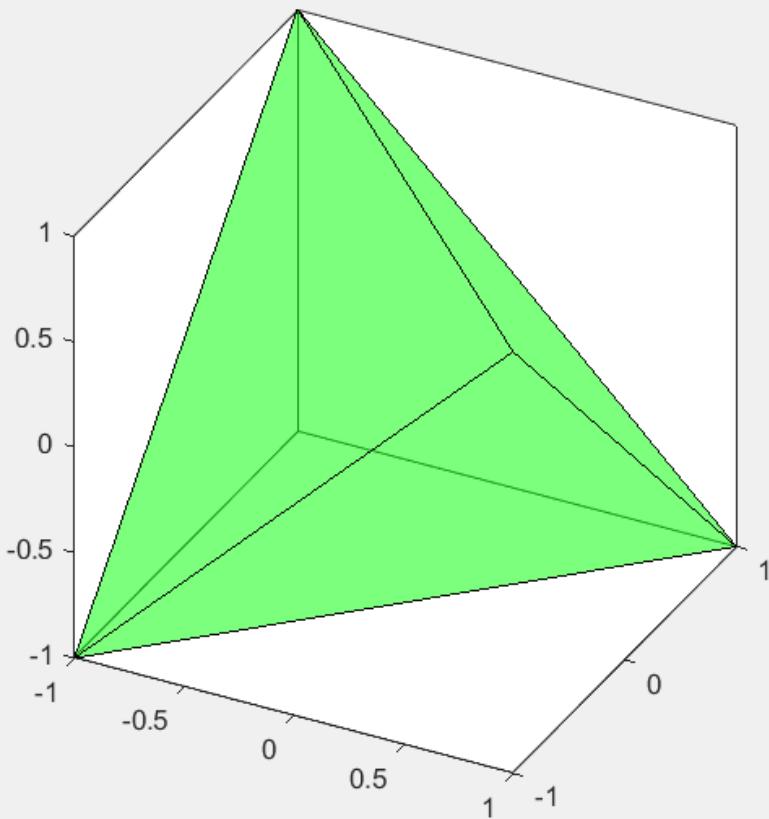
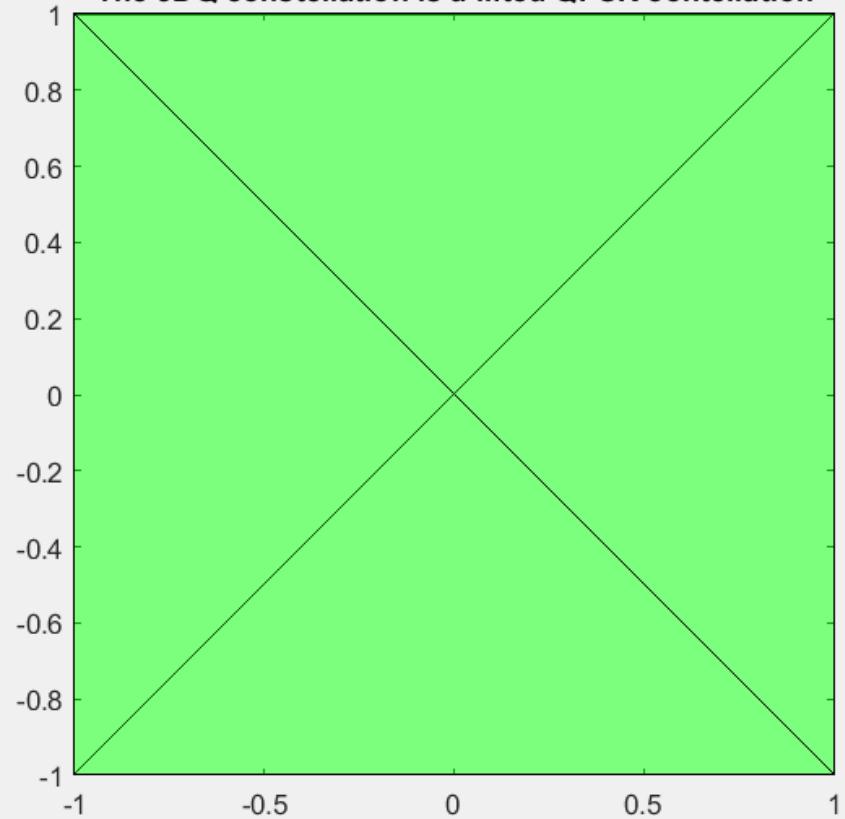


The 3DQ constellation: Vertices of a tetrahedron!



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

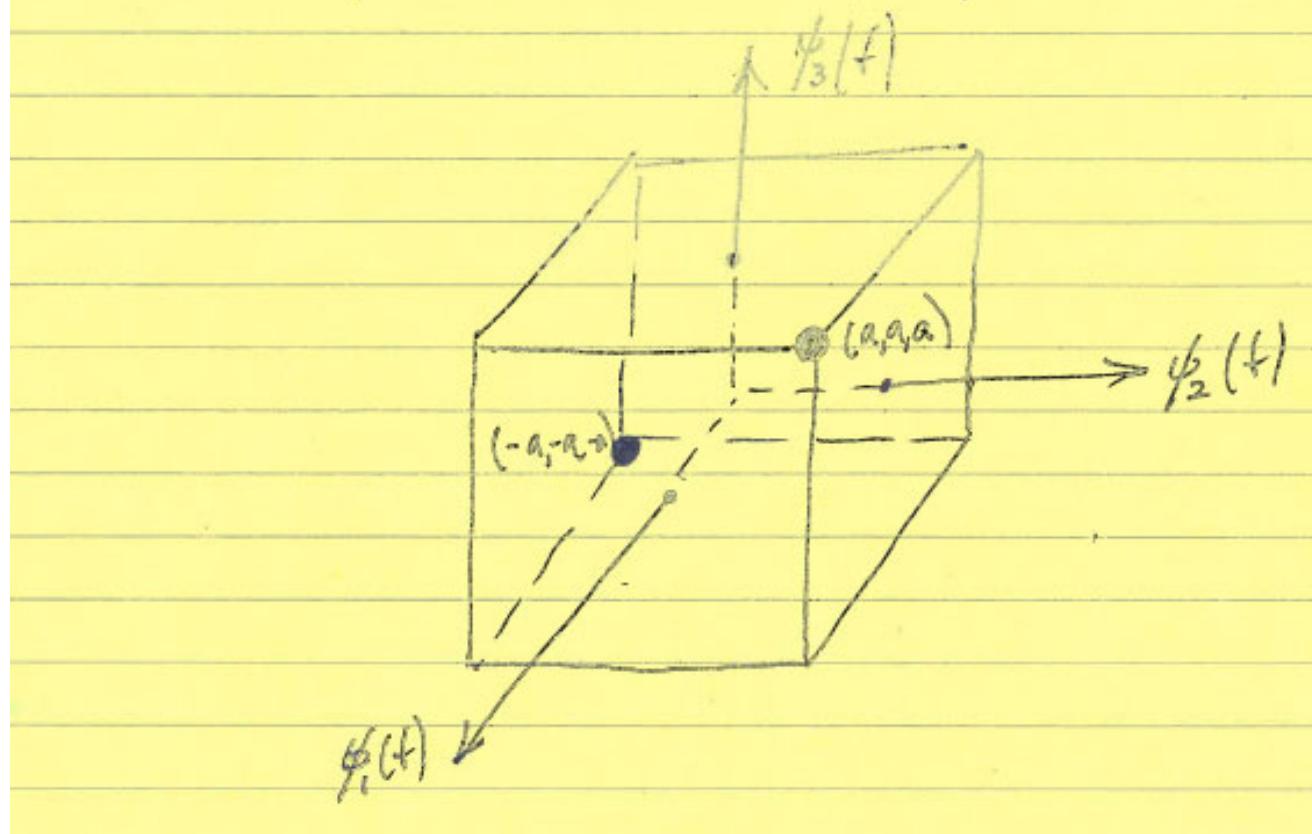
$$2^{n-k} = \sum_{i=0}^1 \binom{n}{i} = 1 + n \quad (2)$$

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

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

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

* Example : If $\bar{C} = (0 \ 0 \ 1)$ and $\bar{F} = (1 \ 1 \ 0)$

then $\bar{S} = \bar{F} \bar{H}^T$

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

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

From table

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

Hamming $(7, 4, 3)$ code:

$$H = \left[\begin{array}{cccc|cc} 1 & 0 & 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 & 1 \end{array} \right] = [P^T | I_3]$$

All nonzero column vectors of 3 bits

Therefore

$$G = [I_4 | P] = \left[\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{array} \right]$$

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

$$\bar{C} = (0110) \left[\begin{array}{c|cc} G & = & (0100011) \\ \hline & + & (0010101) \end{array} \right] = \underline{\underline{(0110110)}}$$

Suppose error in 4th position, so that

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

- ① Compute syndrome:

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

Homework:
Write as equations/circuit

- ② This corresponds to the 4th column of H .
 ↑ transpose of the
 ∴ Error in 4th position.

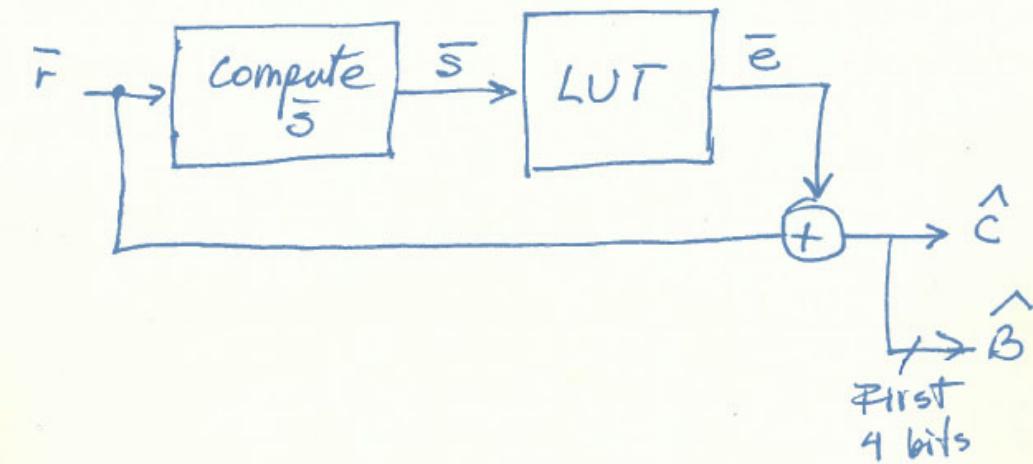
$$\hat{a} = (0 \ 1 \ 1 \ \underset{\substack{\uparrow \\ \text{Information} \\ \text{bits}}}{0} \ 1 \ 0) \rightarrow (0 \ 1 \ 1 \ 0) = \hat{b}$$

Look-up Table :

Syndrome \bar{s}	Error vector \bar{e}
0 0 0	0 0 0 0 0 0 0
1 1 0	1 0 0 0 0 0 0
0 1 1	0 1 0 0 0 0 0
1 0 1	0 0 1 0 0 0 0
1 1 1	0 0 0 1 0 0 0
1 0 0	0 0 0 0 1 0 0
0 1 0	0 0 0 0 0 1 0
0 0 1	0 0 0 0 0 0 1

H^T

HD 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

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

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

where Y_i denote MF outputs

Soft-decision decoding

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

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

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

EE161

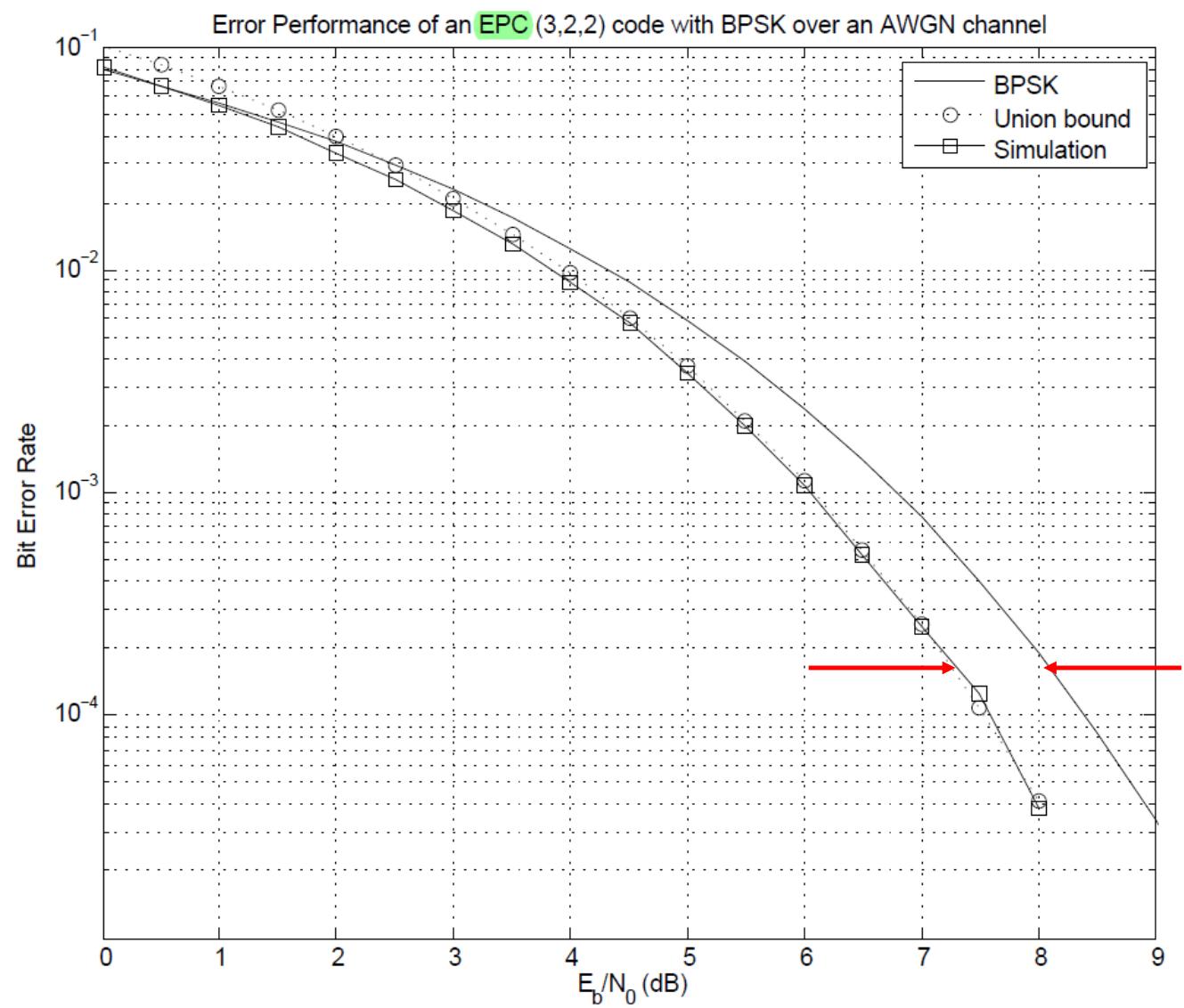
CORRELATIONS OF (3,2,2)
EPC CODE

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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	a - a - a	$Y_1 - Y_2 - Y_3$	
+ 1	+ 1 0	0 - a - a a	$-Y_1 - Y_2 + Y_3$	
1 0	+ 0 1	- a a - a -	$-Y_1 + Y_2 - Y_3$	$a = \sqrt{E_s / 3}$

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

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	-a + a + a	$-Y_1 + Y_2 + Y_3$
+ 1	+ 1 0	+ a - a - a	$+Y_1 + Y_2 - Y_3$
1 0	+ 0 1	+ a - a + a	$+Y_1 - Y_2 + Y_3$

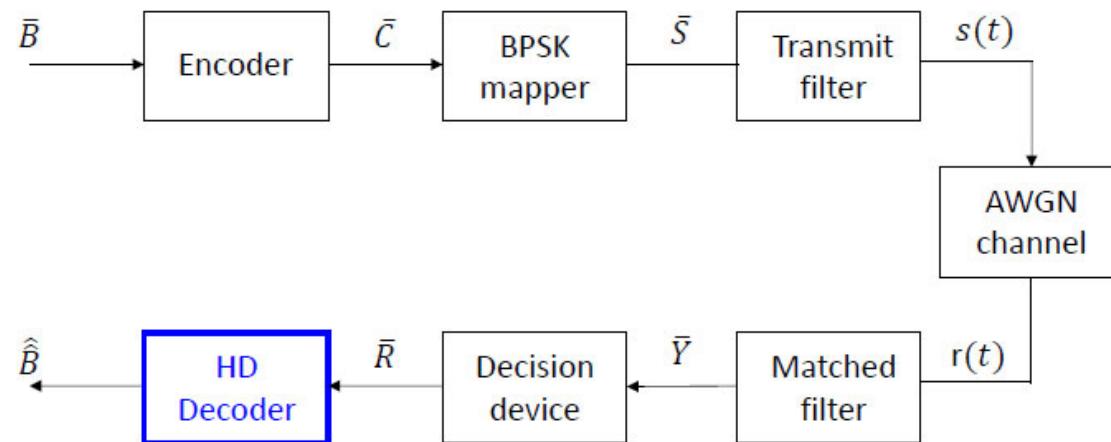


Hard-decision (HD) decoding versus soft-decision (SD) decoding

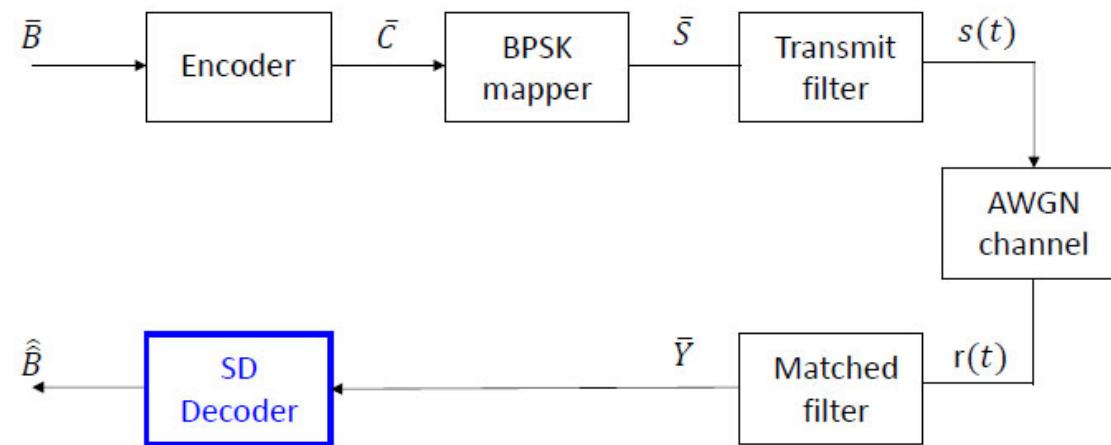
- HD decoding: Correct errors after ML estimation (decision device), R_k
- SD decoding: Use of matched-filter outputs, Y_k

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

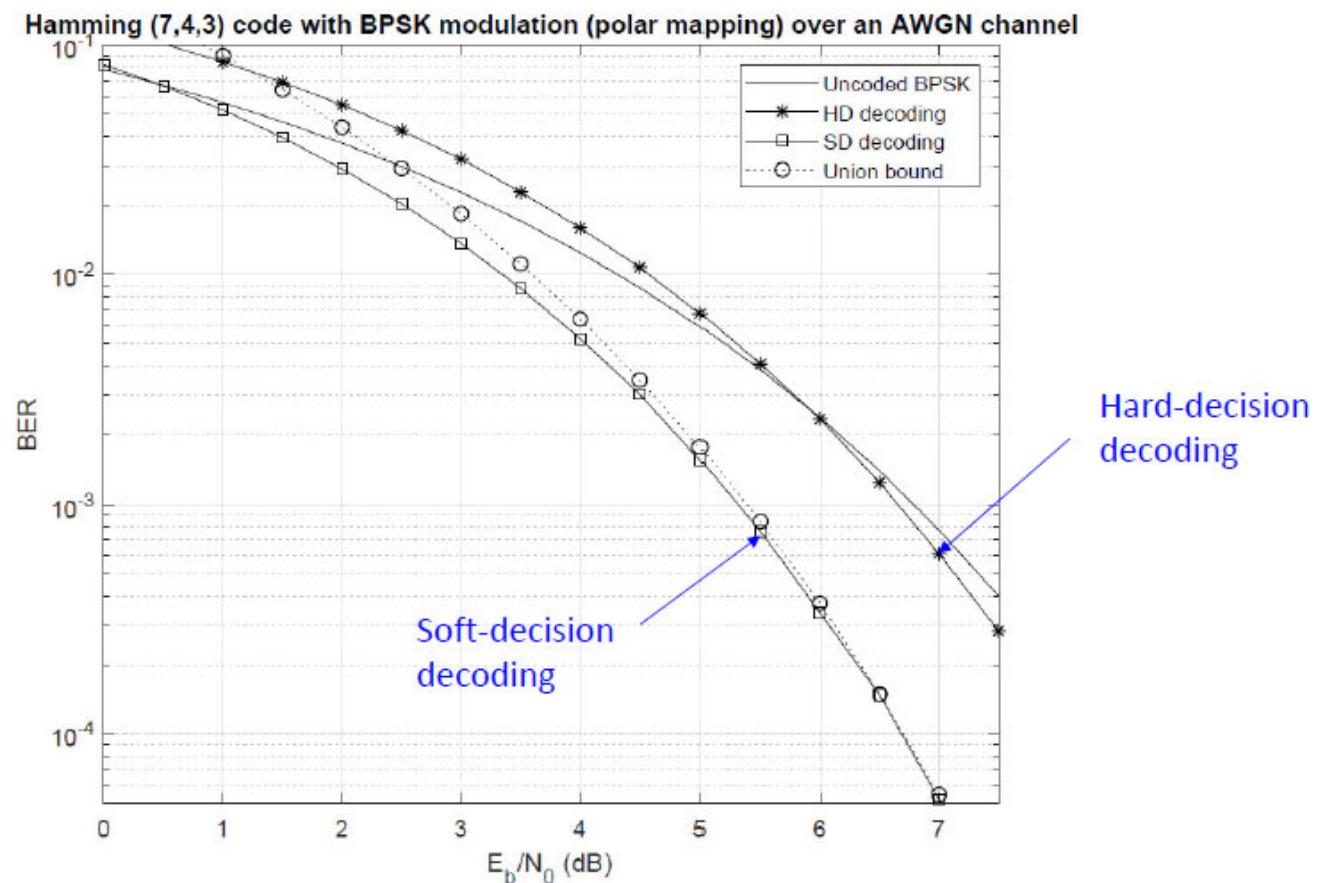
Hard-decision decoding



Soft-decision decoding



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



Coding gain and “real” (effective) coding gain

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

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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 $E_s = k \underbrace{E_b}_{\uparrow}$

or

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

The ^{min. Euclidean} distance between k codewords modulated is $d_{\min}^E = \sqrt{4d_{\min} \cdot a^2}$
note, this is the pulse energy

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

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

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

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

(dB)

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