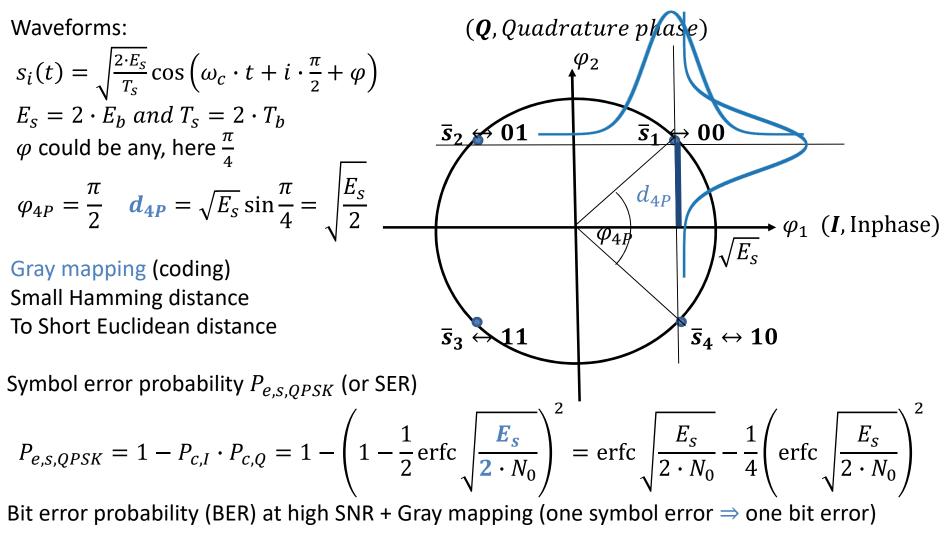
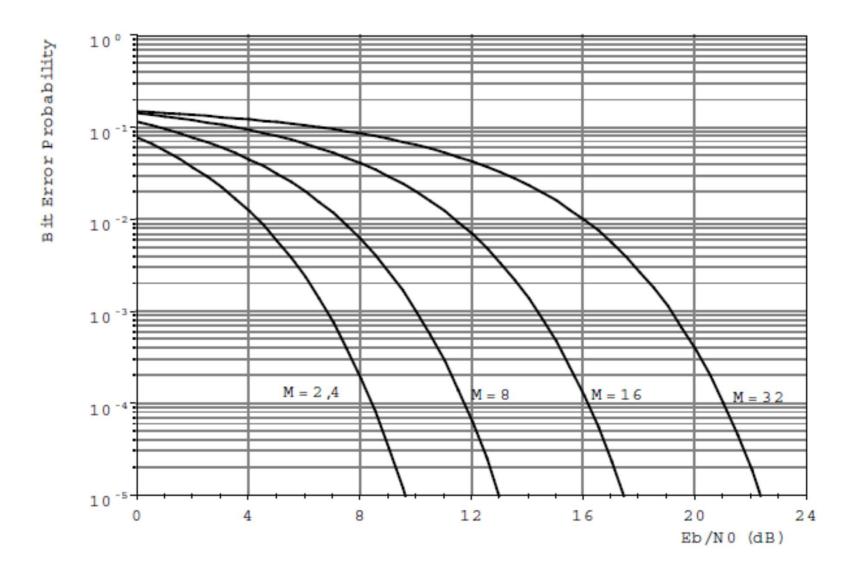
Űrkommunikáció Space Communication 2023/12.

Quaternary (Quadrature) Phase Shift Keying – QPSK

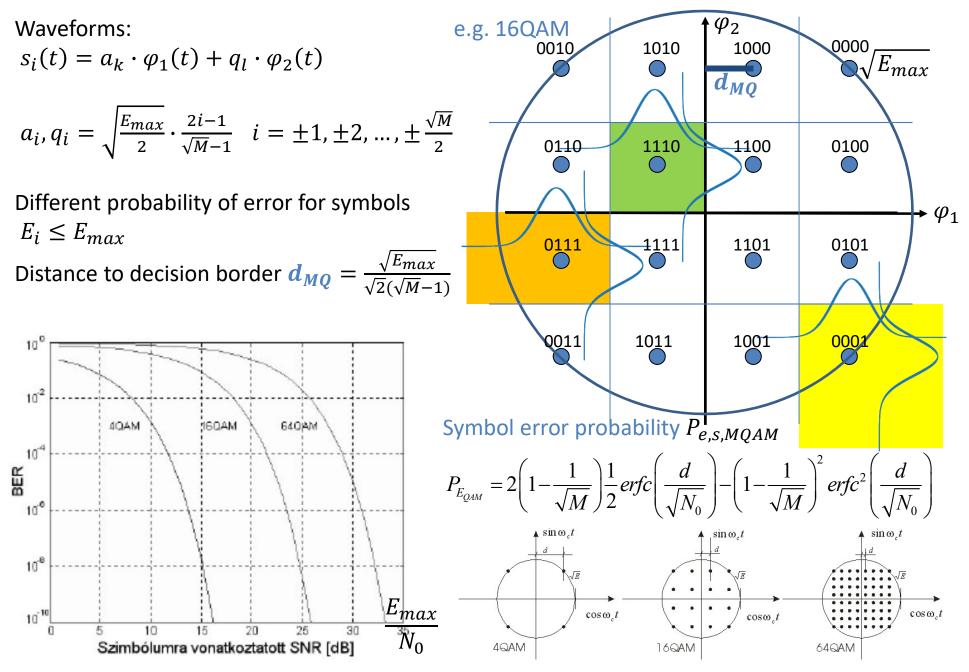


$$P_{e,b,QPSK} \cong \frac{1}{2} \operatorname{erfc} \sqrt{\frac{2 \cdot E_b}{2 \cdot N_0}} - \frac{1}{4} \left(\operatorname{erfc} \sqrt{\frac{E_s}{2 \cdot N_0}} \right)^2 = \frac{1}{2} \operatorname{erfc} \sqrt{\frac{E_b}{N_0}} = P_{e,BPSK}$$

M-ary Phase Modulation



M-ary Quadrature Amplitude Modulation – MQAM



MPSK versus MQAM

BPSK **MPSK** MQAM Symbol energy: $E_s = P_{MP} \cdot T_b \cdot ld M$ $E_{max} = P_{MO} \cdot T_b \cdot ld M$ $E_b = P_{BP} \cdot T_b$ Euclidean Distance to decision border: $d_{MP} = \sqrt{E_s} \sin \frac{\pi}{M} \qquad \qquad d_{MQ} = \frac{\sqrt{E_{max}}}{\sqrt{2}(\sqrt{M}-1)}$ $d_{BP} = \sqrt{E_b}$ Same Euclidean Distance to decision border ↔ Approximately the same probability of error $\frac{d_{MP}}{d_{BP}} = 1 = \frac{\sqrt{E_s} \sin \frac{\pi}{M}}{\sqrt{E_b}} = \frac{\sqrt{P_{MP} \cdot T_b} \cdot ld M \sin \frac{\pi}{M}}{\sqrt{P_{BP} \cdot T_b}} \leftrightarrow \frac{P_{MP}}{P_{BP}} = \frac{1}{\left(\sin \frac{\pi}{M}\right)^2 \cdot ld M}$ $\frac{d_{MQ}}{d_{RP}} = 1 = \frac{\sqrt{E_{max}}}{\sqrt{2}(\sqrt{M}-1)} = \frac{\sqrt{P_{MQ} \cdot T_b \cdot ld M}}{\sqrt{2}(\sqrt{M}-1)\sqrt{P_{RP} \cdot T_b}} \leftrightarrow \frac{P_{MQ}}{P_{RP}} = \frac{2 \cdot (\sqrt{M}-1)^2}{ld M}$

n=ld M bit	$M = 2^n$	$B_n \cdot T_b$	P_{MP}/P_{BP} [dB]	P_{MQ}/P_{BP} [d	B] Comments
1	2	1	0	0	
2	4	0,5	0	0	QPSK! B/2
3	8	0,33	3,57	x	
4	16	0,25	8,17	6,53	QAM>PSK
5	32	0,2	13,2	x	
6	64	0,17	18,4	12,1	6 dB for QAM

M-ary orthogonal Frequency Shift Keying – MFSK

Waveforms:

$$s_i(t) = \sqrt{\frac{2 \cdot E_s}{T_s}} \cos[(\omega_c + i \cdot \delta \omega) \cdot t + \varphi_i] \quad i = 0, \dots, M - 1$$

Frequency deviation: $\delta \omega$ φ_i could be any

Signal set is orthogonal if $\delta \omega = 2\pi/T_s$

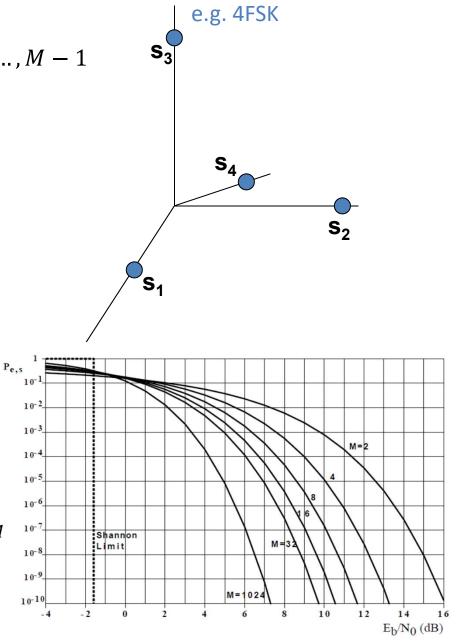
Symbol error probability

$$P_E = \frac{M-1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} \Phi\left(u - \sqrt{\frac{2E}{N_0}}\right) \Phi^{M-2}(u) e^{-u^2/2} du$$

where $\Phi(u)$ is cumulative distribution of
 $G(\mu = 0, \sigma = 1)$ standard Gaussian function
 $\Phi(u) = \frac{1}{2\pi} \int_{-\infty}^{u} e^{-x^2/2} dx$

Upper bound at high SNR

$$P_{E} \leq \begin{cases} 1, & E/N_{0} \leq \ln M \\ exp\left\{-\left(\sqrt{E/N_{0}} - \sqrt{\ln M}\right)^{2}\right\}, & \ln M < E/N_{0} \leq 4 \cdot \ln M \\ exp\{-(E/2N_{0} - \ln M)\}, & E/N_{0} > 4 \cdot \ln M \end{cases}$$



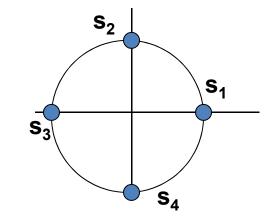
M-ary Biorthogonal Frequency Shift Keying

e.g. biorthogonal 4FSK

M -ary *biorthogonal* signal constellation is obtained from a set of M/2 orthogonal signals (such as M/2-*FSK*)

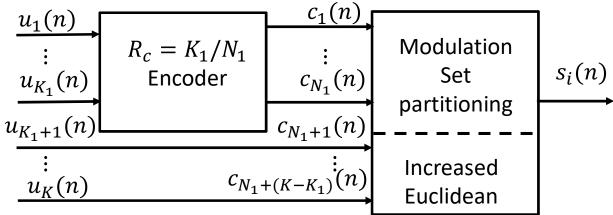
Waveforms: $\pm s_i(t) \ i = 0, ..., M/2$

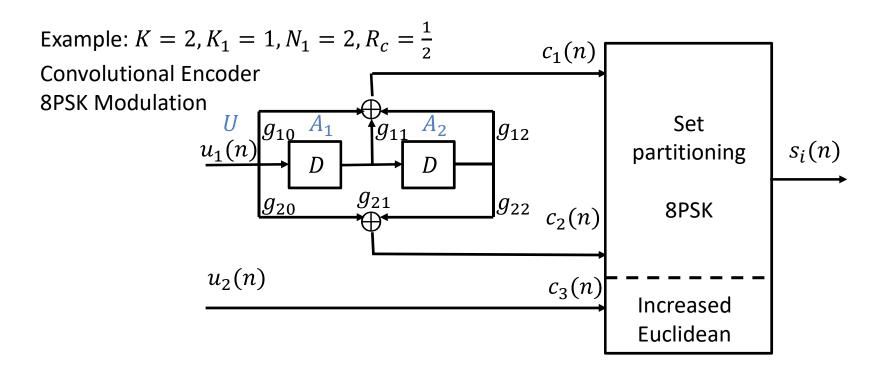
Symbol error performance is similar as by QPSK

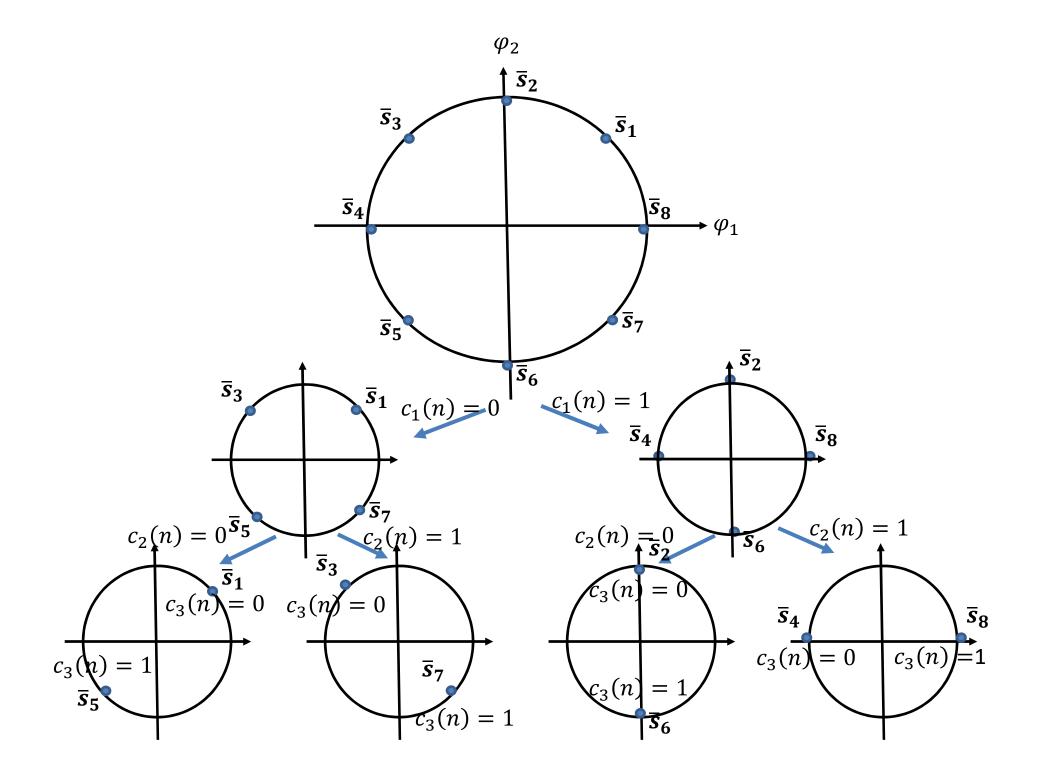


Coded Modulation

Goal: Increasing Euclidean distance using set partitioning. Redundant modulation symbols. Advantage at Demodulation: Protection of subset selection by encoding, increased distance inside the subset.



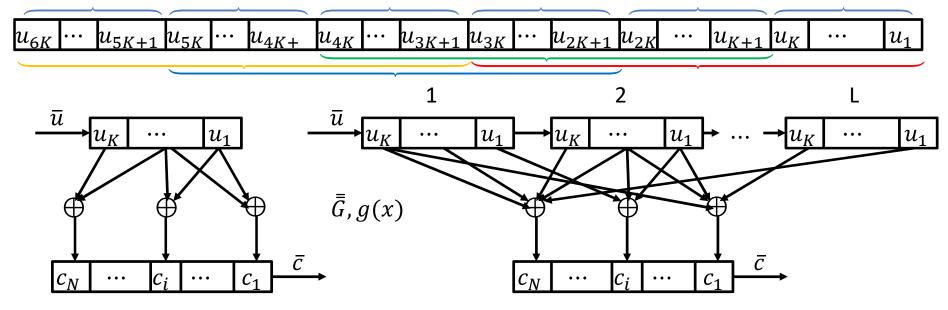




Convolutional coding, trellis coding, Viterbi (de)coding

Block coding versus convolutional coding: both linear, but convolutional coding have memory Block diagram of code generation with shift registers:

Block coding (fixed window) Convolutional coding (moving window)



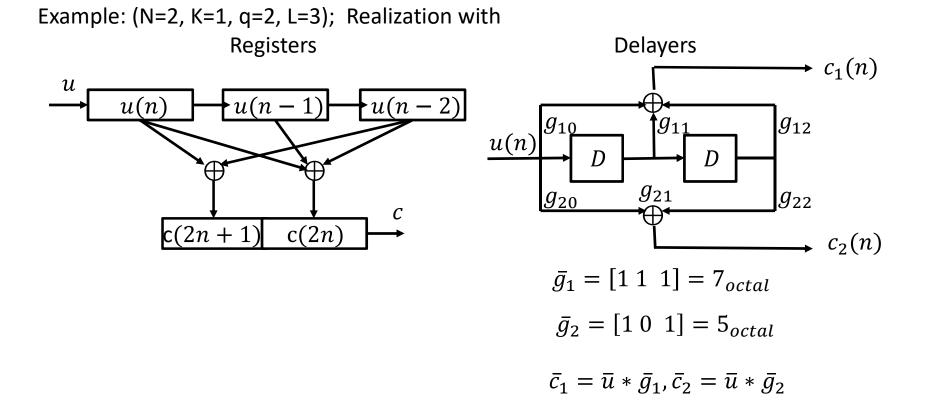
Parameters of convolutional codes: N, K, L, q, $R_c = K/N$

- N: Length of the code vector
- K: Length of the message vector
- L: Constrain length
- *R_c*: Coding rate
- q: Size of GF(q), mostly binary, q=2

Convolutional coding

Some features:

- Linear codes, because remains in the code space
- Generator polynomials are non-trivial, computer search for appropriate polynomials are necessary; there are no algorithms for defining the polynomials
- There are systematic, but usually non-systematic codes
- Possibility for "soft decoding"



Convolutional coding

Generates the code vectors as discrete convolution of the message vector with the generator vectors:

$$\bar{c}(n) = [c_1(n), c_2(n), \cdots, c_N(n)],$$
$$c_i(n) = \left(\sum_{j=0}^{L-1} u(n-j) \cdot g_{ij}\right) \mod q$$

Unilateral Z-transform: converts a discrete-time signal x(n), which is a sequence of real or complex numbers, into a complex frequency-domain representation X(z).

$$X(z) = \mathbb{Z}\{x(n)\} = \sum_{n=0}^{\infty} x_n \cdot z^{-n}$$

In our case discrete –time sequence (i.e. u(n), c(n), etc.) of symbols (i.e. GF(q) elements) into frequency domain. Let use the notation: $D = z^{-1}$

Transformation of message sequence u(n) into message polynomial U(D)

$$U(D) = \mathbb{Z}\{u(n)\} = \sum_{j=0}^{j} u_j \cdot D^{j}$$

Transformation of generator sequence $g_i(n)$ into generator polynomial $G_i(D)$

$$G_i(D) = \sum_{j=0}^{L-1} g_{ij} \cdot D^j$$

Convolutional coding

A code polynomial

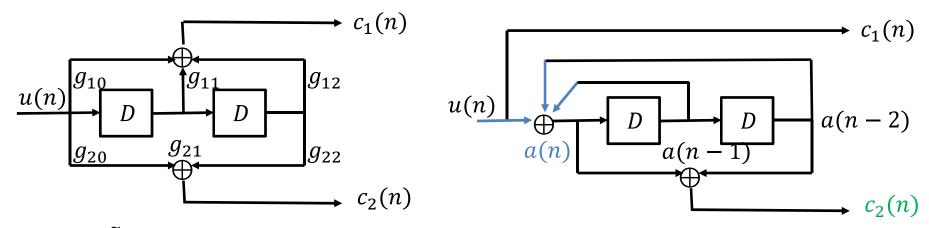
 $C_i(D) = U(D) \cdot G_i(D)$

The code vector polynomial is then the vector of code polynomials $\bar{C}(D) = [C_1(D), C_2(D), \cdots, C_N(D)] = U(D) \cdot \bar{G}^T(D)$

Where $\overline{G}^{T}(D)$ is the transpose of the vector of generator polynomials $G(D) = [G_1(D), G_2(D), \dots, G_N(D)]$ + $c_1(n)$ Example: (N=2, K=1, q=2, L=3); $G_1(D) = g_{10} + g_{11} \cdot D^1 + g_{12} \cdot D^2 = 1 + D^1 + D^2$ g_{12} g_{10} g_{11} $G_2(D) = g_{20} + g_{21} \cdot D^1 + g_{22} \cdot D^2 = 1 + D^2$ u(n)Let the message: u(n) = 10011..... g_{21} g_{22} g_{20} $U(D) = 1 + D^3 + D^4 + \cdots$ Z-transform: ÷ $\rightarrow c_2(n)$ Code polynomials: $C_1(D) = U(D) \cdot G_1(D) = 1 + D^3 + D^4 + D + D^4 + D^5 + D^2 + D^5 + D^6 + \cdots$ $= 1 + D + D^{2} + D^{3} + D^{6} + \cdots$ $C_2(D) = U(D) \cdot G_2(D) = 1 + D^3 + D^4 + D^2 + D^5 + D^6 + \cdots$ $= 1 + D^{2} + D^{3} + D^{4} + D^{5} + D^{6} + \cdots$ $C(D) = [C_1(D), C_2(D)]$ D^0 D^1 D^2 D^3 D^4 D^5 D^6 The code sequence: $c(n) = \overrightarrow{11} \quad \overrightarrow{10} \quad \overrightarrow{11} \quad \overrightarrow{01} \quad \overrightarrow{01} \quad \overrightarrow{01} \quad \overrightarrow{11} \dots$

Systematic versus non-systematic codes

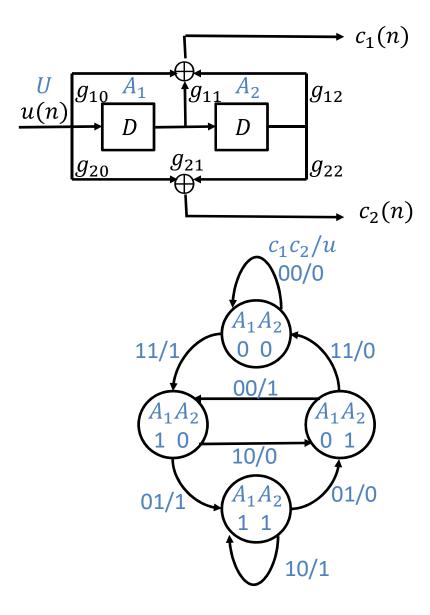
- NSC: Non-recursive, non-Systematic, Convolutional codes The current state of the coder is independent from the structure Usual case except turbo coding, where RSC is applied
- RSC: Recursive, Systematic Convolutional codes
 The current state of the coder depends from the structure
 NSC structure
 Derive an RSC from an NSC
 RSC structure



$$\begin{aligned} G_1(D) \Rightarrow \tilde{G}_1(D) &= 1 \text{ Systematic} \\ G_2(D) \Rightarrow \tilde{G}_2(D) &= \frac{G_2(D)}{G_1(D)} = \frac{1+D^2}{1+D^1+D^2} & A(D) = \frac{U(D)}{1+D^1+D^2} \Rightarrow U(D) = A(D) \cdot (1+D^1+D^2) \\ \tilde{C}_1(D) &= U(D) \cdot \tilde{G}_1(D) = U(D) &\Rightarrow u(n) = a(n) + a(n-1) + a(n-2) \\ \tilde{C}_2(D) &= U(D) \cdot \tilde{G}_2(D) = \frac{U(D)}{G_1(D)} G_2(D) &\Rightarrow a(n) = u(n) + a(n-1) + a(n-2) \\ \end{aligned}$$

Graphical representation

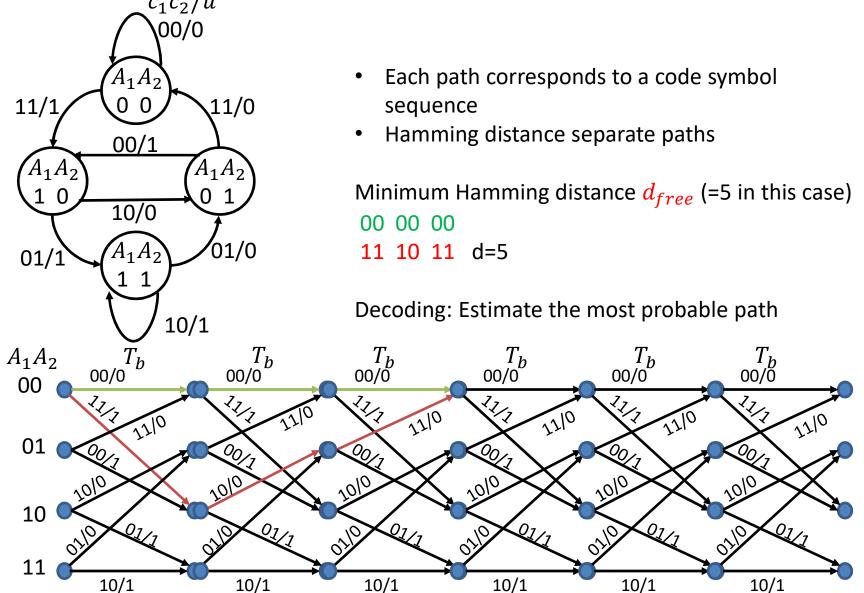
Describing the time behavior of the encoding process: State diagram versus trellis diagram



U	<i>A</i> ₁	A 2	<i>c</i> ₁	<i>C</i> ₂
0	0	0	0	С ₂ 0
	0	0		
1	0	0	1	1
	1	0		
0	0	1	1	1
	0	0		
1	0	1	0	0
	1	0		
0	1	0	1	0
	0	1		
1	1	0	0	1
	1	1		
0	1	1	0	1
	0	1		
1	1	1	1	0
	1	1		

Graphical representation

Describing the time behavior of the encoding process: State diagram versus trellis diagram $c_1 c_2/u$



Decoding criterions

• MAP – Maximum A-posteriori Probability criteria

 $\hat{c}(n) = \arg \max_{c(n)} \{ p(c(n) | v(n)) \}; \text{ with } a - posteriori \ p(c(n) | v(n)) \}$ $Or \ equivalently$ $p(\hat{c}(n) | v(n)) \ge p(c(n) | v(n)) \quad \forall \ c(n)$ $Multiplying \ with \ p(v(n))$ $p(\hat{c}(n) | v(n)) \cdot p(v(n)) \ge p(c(n) | v(n)) \cdot p(v(n)) \quad \forall \ c(n)$ According to Bayes criteria $p(\hat{c}(n), v(n)) \ge p(c(n), v(n)) \quad \forall \ c(n)$

 $p(v(n) | \hat{c}(n)) \cdot p(\hat{c}(n)) \ge p(v(n) | c(n)) \cdot p(c(n)) \quad \forall c(n)$

Therefore:

 $\hat{c}(n) = \arg \max_{c(n)} \{ p(v(n) \mid c(n)) \cdot p(c(n)) \}; \text{ with } a - priori p(c(n)) \}$

• ML – Maximum Likelihood criteria

$$\hat{c}(n) = \arg \max_{c(n)} \{ p(v(n) \mid c(n)) \}; \text{ without } a - priori \, p(c(n)) \}$$

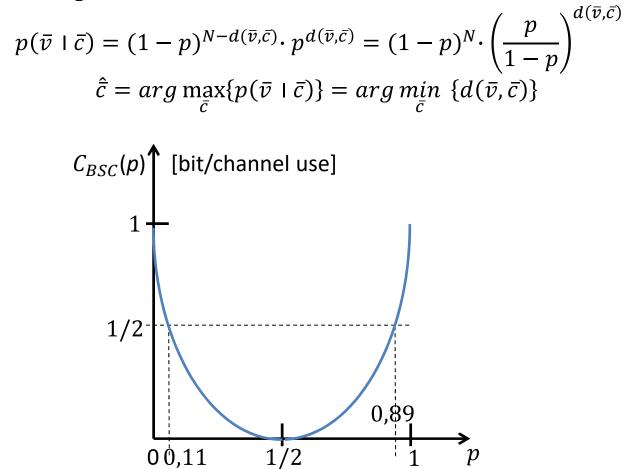
If p(c(n)) uniform, then MAP \equiv ML

Decoding criterions

• Maximum Likelihood estimation over BSC with error probability p

$$\hat{c}(n) = \arg \max_{c(n)} \{ p(v(n) \mid c(n)) \}$$

For vectors of length N



Decoding criterions

• Maximum Likelihood (ML) decision

In favor of the vector \bar{c} from all the possible 2^N vectors (sequence) of length N according to

 $\hat{c} = \arg \max_{\bar{c}} \{ p(\bar{v} \mid \bar{c}) \} = \arg \min_{\bar{c}} \{ d(\bar{v}, \bar{c}) \}$

Number of required distance calculations: 2^N (exponentially increasing).

• Viterbi algorithm, Maximum Likelihood Sequence Estimation (MLSE)

At closing of alternative routes (starting from the same state and ending at the same state of the trellis) discarding path with higher Hamming distance to the received sequence of length N.

Number of required distance calculations: $\sim N$ (Linear increasing).

Example: (N=2, K=1, L=3, q=2), $G_1(D) = 1 + D^1 + D^2$, $G_2(D) = 1 + D^2$

		00 11	
11	11	01 10	
00	01	10	d=5
	CC		

Suppose 00 00 00, decision will be wrong, if we receive 3, 4, or 5 errors, e.g. 10 10 10 -> 11 10 11

Probability of wrong decision at d=5 over BSC(p) $P_{d=5} = \sum_{i=3}^{5} {5 \choose i} \cdot p^{i} \cdot (1-p)^{5-i}$

indifferent

Resulting error probability

• Convolutional coding, Viterbi decoding, BSC(p)

Upper bound of wrong decision probability at d

$$P_{d} = \begin{cases} \sum_{i=\frac{d+1}{2}}^{d} \binom{d}{i} \cdot p^{i} \cdot (1-p)^{d-i} & \text{for } d \text{ odd} \\ \\ \frac{1}{2} \cdot \binom{d}{d/2} \cdot p^{d/2} \cdot (1-p)^{d/2} + \sum_{i=\frac{d}{2}+1}^{d} \binom{d}{i} \cdot p^{i} \cdot (1-p)^{d-i} & \text{for } d \text{ even} \end{cases}$$

Upper bound of resulting error probability at $N \rightarrow \infty$ (at the end of transmission)

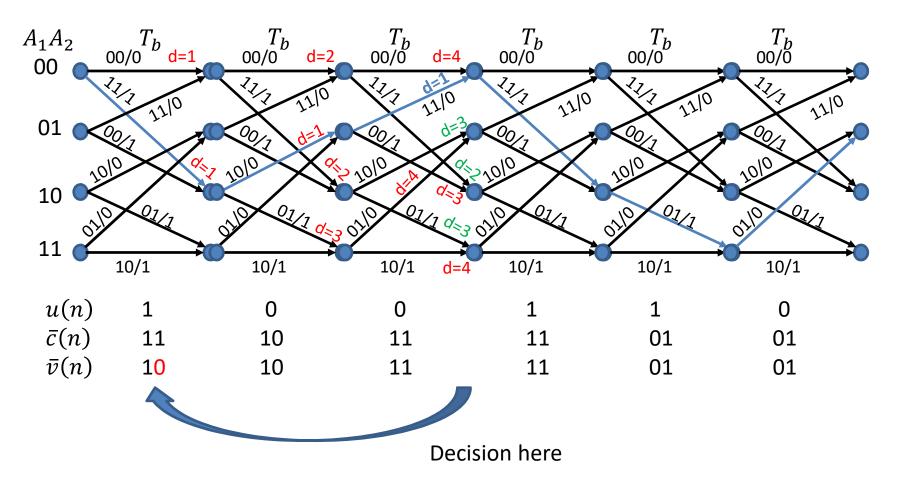
$$P_{e,BSC} < \sum_{d=\frac{d_{free}}{2}}^{\infty} a_d \cdot P_d$$

Where a_d is the number of closing of alternative routes (loops) with Hamming d of constructing paths.

For BPSK over AWGN with SNR $P_e < \sum_{d=d_{free}}^{\infty} erfc \sqrt{SNR \cdot R_c \cdot d}$

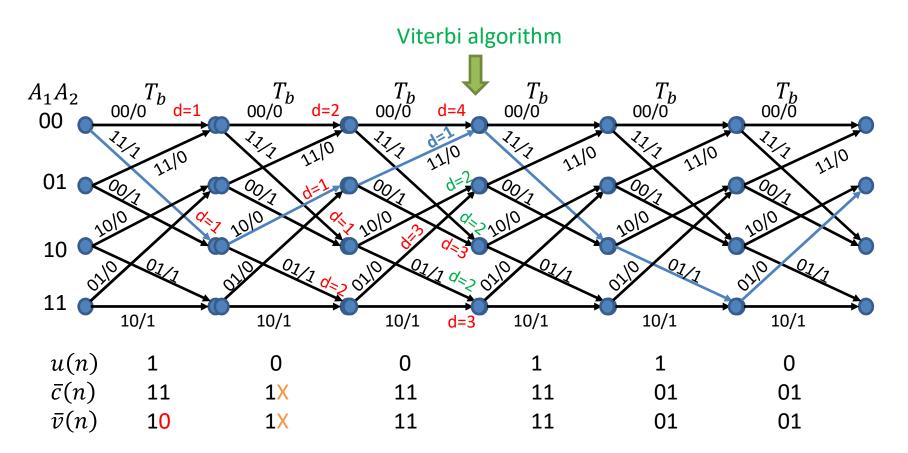
Example: Trellis diagram

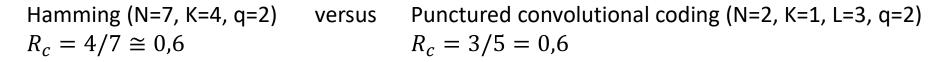
- Example parameter of convolutional coding (N=2, K=1, L=3, q=2)
- Encoding the Message sequence u(n) into Code vector sequence $\overline{c}(n)$
- Cumulate the Hamming distance from the Received vector sequence $\bar{v}(n)$ along all possible path
- Estimate the most probable path

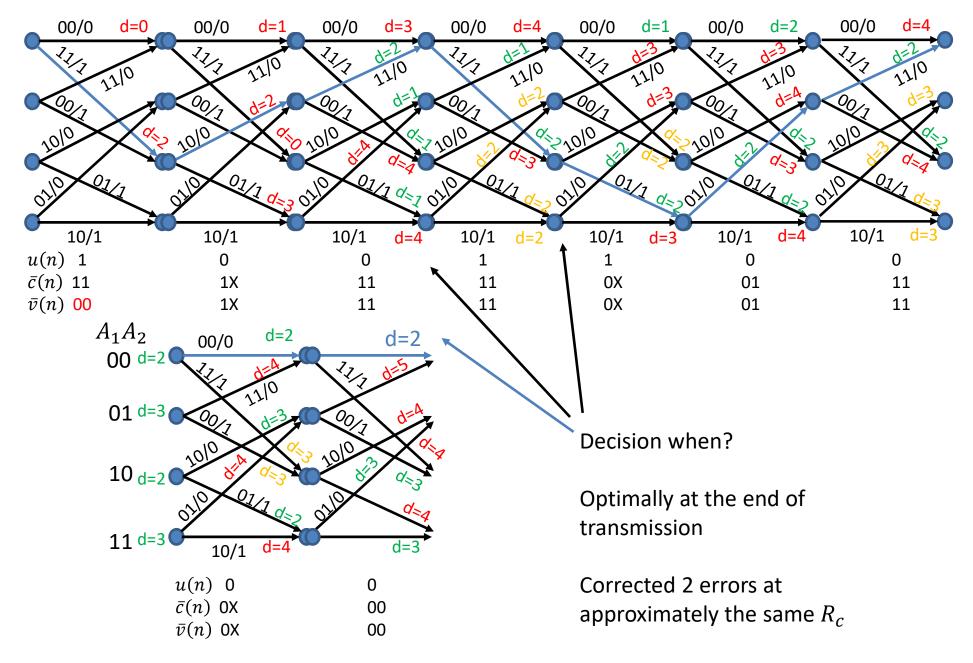


Puncturing

- Indifferent code symbols could be avoided by transmission
- Even other code symbols could be avoided adapting the coding rate R_c to the channel quality







Coded Modulation

Goal: Increasing Euclidean distance using set partitioning. Redundant modulation symbols. Advantage at Demodulation: Protection of subset selection by encoding, increased distance inside the subset.

