

Protection de l'information

Channel coding

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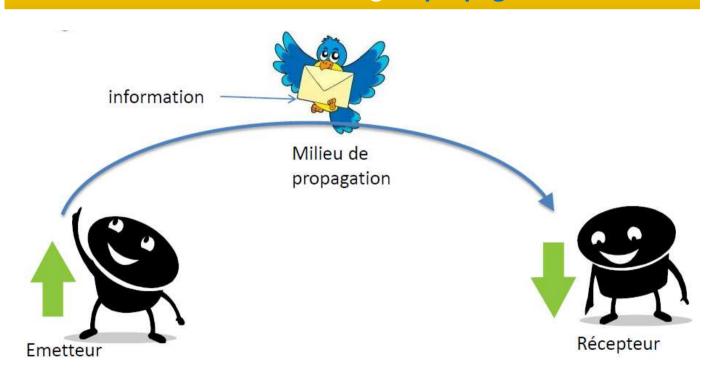


I- Introduction

- 1. History
- 2. Communication Chain

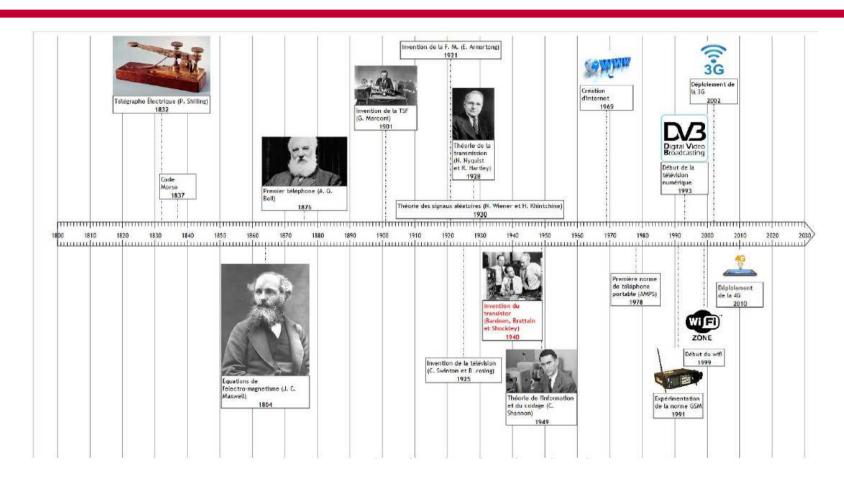
What is a *communication*?

A communication is to transmit an information from a source to a destination through a propagation medium





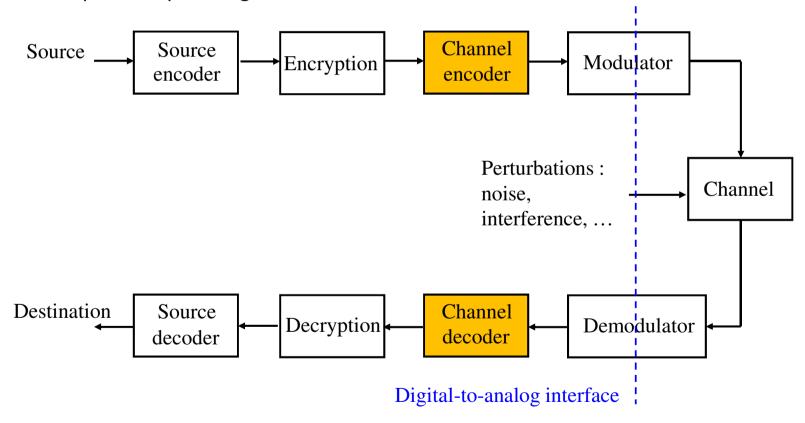
History of communications





Communication chain

Diagram of a point-to-point digital transmission



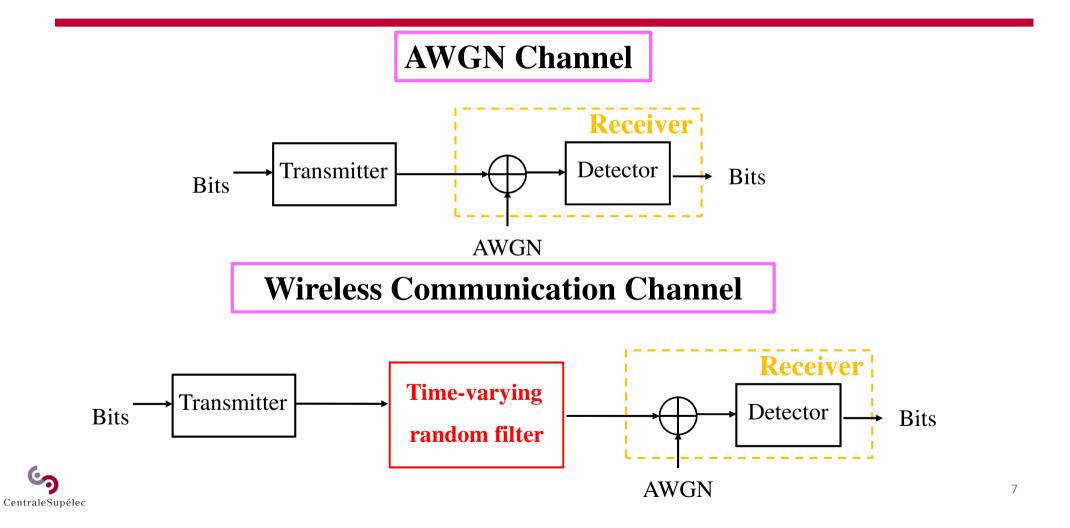




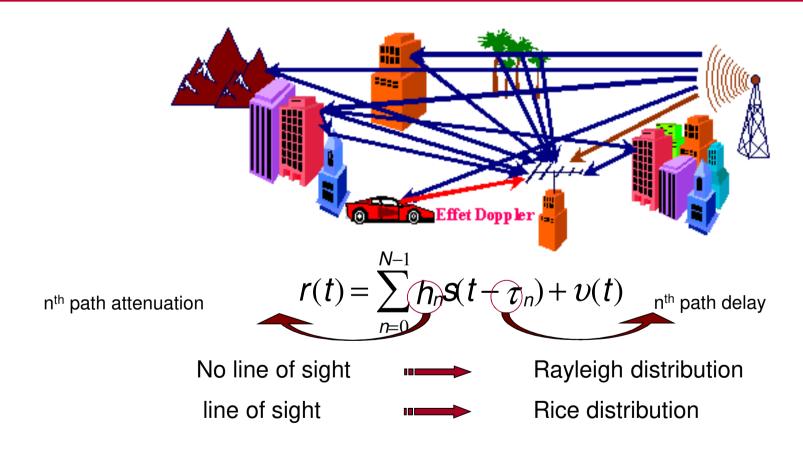
II- Wireless channel

- 1. AWGN channel vs. Wireless channel
- 2. Fading
- 3. Multi-path cannel

AWGN channel vs. Wireless channel



Multi-path channel





Fading distribution

 In a classical environment (theoretical perfect model), noise is generally considered to be AWGN (Additive White Gaussian Noise), modeled by the normal distribution.

$$f_h(h) = \frac{e^{\frac{h^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}}$$

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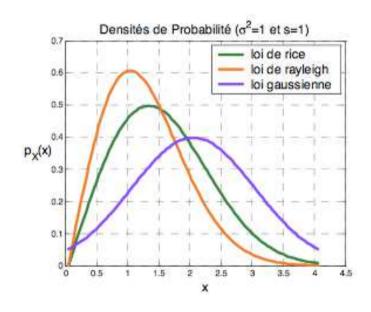
■ In the case of a fading channel, the probability density of the attenuations will follow a law of the form:

$$f_h(h) = \frac{h}{\sigma^2} e^{-\left(\frac{h^2 + A^2}{2\sigma^2}\right)} I_0\left(\frac{hA}{\sigma^2}\right)$$
 Rice distribution

- If A=0, we come to the Rayleigh distribution
- Rayleigh distribution is the preferred fading model because it models the most severe conditions
 with the simplest expression.

Rice vs. Rayleigh

- The Rayleigh model is used more than Rice model for the following reasons:
 - The Rayleigh model corresponds to a propagation without line of sight, and therefore more constraining. This allows to rely on the worst case
 - The Rayleigh model corresponds to many practical cases of propagation, in an urban environment or in indoor propagation (indoor)
 - The mathematical expression of a Rayleigh distribution is simpler than that of a Rice one





Time variations of the channel: Coherence time

- The coherence time T_c gives an approximation of the time during which the behavior of the channel is relatively constant
- The coherence timeT_c allows to characterize the time variation of the channel in the <u>time domain</u>
- Principle : compare T_c to T, symbol period.

	Fading variation
$T_c > T$	Slow fading
$T_{c} < T$	Fast fading



Slow fading vs. fast fading

Slow fading :

- The channel changes but *slowly*
- Channel coefficients remain constant all the time of a frame transmission

$$r(t) = \sum_{n=0}^{N-1} h_n s(t - \tau_n) + v(t)$$

- Fast fading :
 - The channel changes very quickly
 - It is impossible to consider the gains of the paths as constant on an observation window

$$r(t) = \sum_{n=0}^{N-1} h_n(t) s(t - \tau_n(t)) + v(t)$$





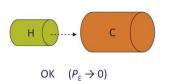
III- Compensation

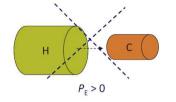
1. Information theory review: channel capacity

« The theory concerned the performance of the very best system possible and how to approach that performance (without explicitely designing the system). An understanding of each piece of the system was necessary to achieve this objective. »

2. Channel coding principle and classification

Channel capacity



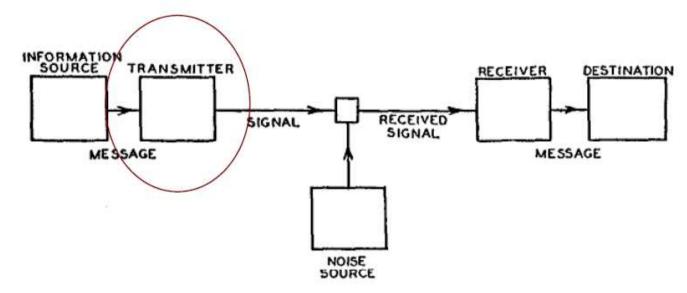


- It is the maximum data rate that may be carried under certain conditions on a transmission channel with an error probability tending to zero
 - The maximization is over all sources which might be used as input to the channel
- Concept closely related to the following
 - Bit rate: expressed in bits / s, at which data can be transmitted
 - Bandwidth: frequency band occupied by the transmitted signal
 - Noise: Channel perturbation
 - Error probability
- Target: exploit bandwidth as efficiently as possible
- Interpretation : If Di <C, there is at least one code allowing to have an error-free transmission.



Channel capacity: the under hidden property







« Choosing the encoding relationship between the source output and the channel input is the <u>most important degree of freedom</u> that we have in designing a system <u>for reliable communication</u>. »



Nyquist theorem

- For bandwidth B, the highest practical rate is 2B bits/s.
- With a multilevel signal, where M is the number of discrete elements, Nyquist's theorem becomes:

$$C = 2B \log_2 M$$

Be careful to the overload for the receiver!



Shannon Theorem

- The higher the data rate, the more disruptive random noise will be detrimental
- Key parameter: Signal-to-noise ratio

$$SNR_{dB} = 10\log_{10} \frac{\text{Puissancedu signal}}{\text{Puissancedu bruit}}$$

■ Gaussian channel capacity (resources are bandwidth B and SNR) in bits/s:

$$C = B \log_2(1 + SNR)$$



Coding theory



Irving Reed



Reed-Muller (1954)

Codes de



Richard Hamming



Marcel Golay

Code de Golay (1949)





David Muller

Codes convolutifs

Codes produits

(1954)

(1968)

Codes LDPC

(1960)



Codes concaténés (1965)

Code de

(1947)

Hamming





Algorithme de Viterbi (1967)

Andrew Viterbi



Peter Elias



Robert Gallager

Codes Reed-

Solomon (1960)





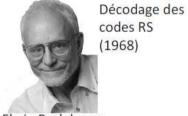
Modulations codées en treillis (1978)

Gottfried Ungerboeck





Irving Reed Gustave Solomon



Elwin Berlekamp



Claude Berrou



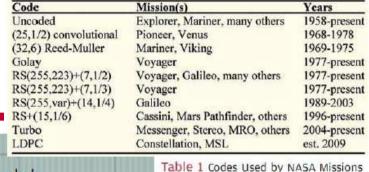
Alain Glavieux

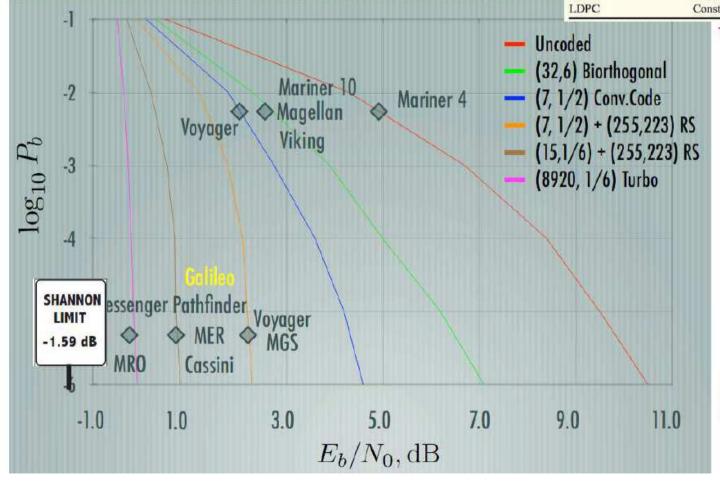




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60 years of coding...





R. J. McEliece, « Viterbi's impact on the exploration of the solar system », 2005



Minimum Distance

- The Hamming distance between two binary codewords: number of different bits. For example, d(0001, 0111) = 2
- Minimum distande of a code $d_{\min} = \min_{\mathbf{y}_1, \mathbf{y}_2; \mathbf{y}_1 \neq \mathbf{y}_2} d(\mathbf{y}_1, \mathbf{y}_2)$
- Since the sum of two codewords is another codeword, and the distance between two codewords is the weight of their sum, we also have:

$$d_{\min} = \min_{\mathbf{y}; \mathbf{y} \neq \mathbf{0}} P(\mathbf{y})$$

- Determining d_{\min} for block codes
 - If the number of codewords is not too high: exhaustive search
 - Otherwise : d_{min} = number of linearly dependent (zero sum) columns of **H** (generator matrix).



Minimum Distance

$$d(a, b) = d(b, a)$$
 Symmetry

$$d(a,b) \ge 0$$
 Positivity

$$d(a, b) = 0$$
 if and only if $a = b$

$$d(a,b) + d(b,c) \ge d(a,c)$$
 Triangular inequality

$$d(a \oplus x, b \oplus x) = d(a, b)$$
 Invariance



Minimum Distance: fundamental parameter

Detect an error

✓ How many errors can be detected with certainty ? d-1

The receiver definitely detects ALL false messages, as long as the number of errors N verifies 0 <N <d. Some false messages with d errors (or more) are not detected.

Correct an error

✓ How many errors can be corrected with certainty? $\lfloor \frac{d-1}{2} \rfloor$

The messages are corrected as long as the number of errors N satisfies $0 \le N < d/2$ (strict). False messages such as $d/2 \le N$ are not always well corrected.

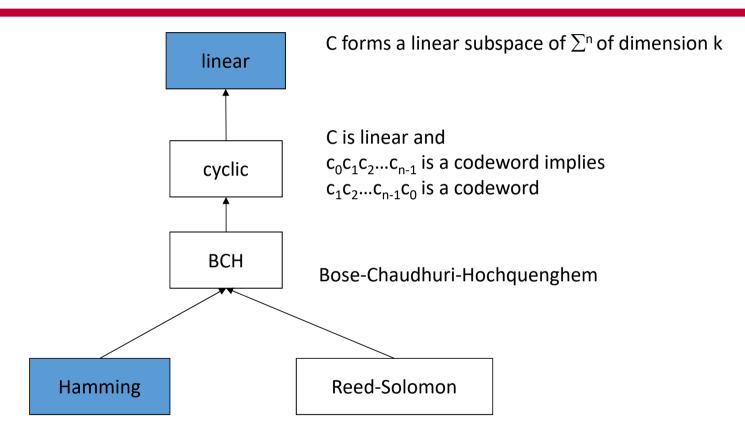


Coding - Examples

- Examples of linear block codes
 - Parity codes
 - Repetition code
 - Hamming code
- Cyclic codes
 - We study here a subclass of the linear codes defined on GC(2).
 - For each codeword of length *n*, a polynomial of degree is associated with the most coefficients equal to the number of bits of the codeword.
 - Cyclic codes are mainly used for error detection.
 - Definition : A linear code C(n, .) is cyclic if any cyclic permutation of a codeword is also a codeword.



Hierarchy of codes







Block linear code

- A linear code is simply defined by a generator matrix.
- Message x generator matrix = coded message, for example:

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

■ So the matrix $\begin{pmatrix} 0 & 0 & 1 \\ 1 & 1 & 1 \end{pmatrix}$ generates the code:

Information bits	00	01	10	11
Corresponding codeword	000	111	001	110



Systematic block linear code

- A systematic codeword contains the information bits as a prefix.
- For this case, the generator matrix contains an identity block on the left.
- Example of the systematic linear code of even parity:

$$\begin{pmatrix} 0 & 0 \\ 0 & 1 \\ 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{1} & \mathbf{0} & 1 \\ \mathbf{0} & \mathbf{1} & 1 \end{pmatrix} = \begin{pmatrix} \mathbf{0} & \mathbf{0} & 0 \\ \mathbf{0} & \mathbf{1} & 1 \\ \mathbf{1} & \mathbf{0} & 1 \\ \mathbf{1} & \mathbf{1} & 0 \end{pmatrix}$$

■ In block representation, it gives M * (I H) = (M C)



Error detection – Syndrom method

- 1. We receive the coded message with a possible error R = M * (I H) + E.
- 2. We multiply by the control matrix $\begin{pmatrix} -H \\ I \end{pmatrix}$

$$R*\begin{pmatrix} -H\\I \end{pmatrix} = M*\begin{pmatrix} I&H \end{pmatrix}*\begin{pmatrix} -H\\I \end{pmatrix}+E*\begin{pmatrix} -H\\I \end{pmatrix}$$

3. Important property for decoding in the case where the H block is square

$$\begin{pmatrix} I & H \end{pmatrix} * \begin{pmatrix} -H \\ I \end{pmatrix} = 0$$

4. By the preceding property, we obtain the syndrome: $S = R * \begin{pmatrix} -H \\ I \end{pmatrix} = E * \begin{pmatrix} -H \\ I \end{pmatrix}$



5. An error is detected if and only if $S \neq 0$.

Cyclic block linear code

- Why use a cyclic code?
 - The problem for all linear codes is to find a good generator matrix.
 - Cyclic codes have a particularly strong structure to help on this point: any cyclic systematic code (n; k) is generated by the polynomials of degree n k dividing $x^n 1$.



Cyclic block linear code: Encoding from a polynomial

1. We start for example from the polynomial

$$x^7 - 1 = (x+1)(x^3 + x + 1)(x^3 + x^2 + 1)$$

- 2. For the codewords of length n = 7, we have 3 possible codes:
 - The one generated by x + 1 encoding messages of length k = n 1 = 6,
 - Those generated by $x^3 + x + 1$ and by $x^3 + x^2 + 1$ encoding messages of length k = n-3 = 4.
- 3. We encode by multiplying the polynomial of the message to be sent by x^{n-k} , then by carrying out the Euclidean division by the generator polynomial g(x). The rest of the division gives the redundancy.
- 4. For example, for polynomial $g(x) = x^3 + x^2 + 1$ and message (1111) $-- \rightarrow x^3 + x^2 + x + 1$: $(x^3 + x^2 + x + 1)x^3 = (x^3 + x^2 + 1)(x^3 + x + 1) + x^2 + x + 1$.
- 5. The encoded message will be (1111111)

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Cyclic block linear code: Error detection with polynomials

- It is possible to obtain a matrix representation but it is more efficient to work with polynomials
 - 1. We have R = M G + E with each row of G being a multiple of g (x). So M G is a multiple of g (x).
 - 2. The rest of the Euclidean division of R by g(x) is the syndrome.
- For the example of $g(x) = x^3 + x^2 + 1$
 - ✓ Message $M = \begin{pmatrix} 1 & 0 & 1 & 0 \end{pmatrix}$, after encoding $\begin{pmatrix} 1 & 0 & 1 & 0 & 0 & 1 \end{pmatrix}$.
 - ✓ Without error, $r(x) = x^6 + x^4 + 1 = (x^3 + x^2 + 1)g(x)$. The rest is null, so there is no mistake.
 - ✓ With error, the received message is $R = \begin{pmatrix} 1 & 0 & 1 & 0 & 1 \end{pmatrix}$.
 - $r(x) = x^6 + x^4 + x^2 + 1 = (x^3 + x^2 + 1)g(x) + x^2$. The rest is not null, so there is an error.



Particular cyclic block linear code: Hamming code

- Family of cyclic codes with $n = 2^m 1$ and $k = 2^m 1 m = n m$ always having a minimum distance d = 3.
- Perfect code for d = 3, that is, it contains no *unnecessary* redundancy.



Hamming Code

- In the late 1940's Richard Hamming recognized that the further evolution of computers required greater reliability, in particular the ability to not only detect errors, but correct them. His search for error-correcting codes led to the Hamming Codes
- Hamming Codes are still widely used in computing, telecommunication, and other applications.
- Hamming Codes also applied in:
 - Data compression
 - Block Turbo Codes



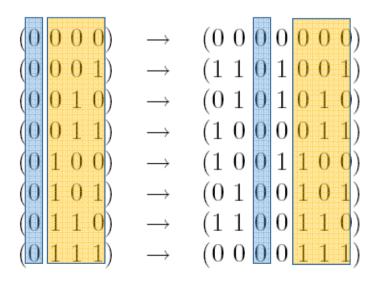
A [7,4] binary Hamming Code (1)

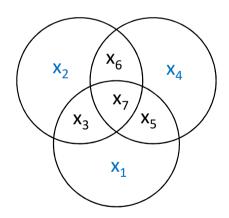
- Let our codeword be $(x_1 x_2 ... x_7) \varepsilon F_2^7$
- x_3 , x_5 , x_6 , x_7 are chosen according to the message (perhaps the message itself is (x_3, x_5, x_6, x_7)).
- $x_4 := x_5 + x_6 + x_7 \pmod{2}$

$$\mathbf{x}_{2} := \mathbf{x}_{3} + \mathbf{x}_{6} + \mathbf{x}_{7}$$

$$x_1 := x_3 + x_5 + x_7$$

Codewords







A [7,4] binary Hamming Code (2)

- Let $a = x_4 + x_5 + x_6 + x_7$ (=1 iff one of these bits is in error)
- Let b = $x_2 + x_3 + x_6 + x_7$
- Let $c = x_1 + x_3 + x_5 + x_7$
- If there is an error (assuming at most one) then abc will be binary representation of the subscript of the offending bit.
- If (y1 y2 ... y7) is received and abc ≠ 000, then we assume the bit abc is in error and switch it. If abc=000, we assume there were no errors (so if there are three or more errors we may recover the wrong codeword).



Parity check matrix

■ A check matrix for a [7,4] binary Hamming Code:

$$L_3 = \left[egin{array}{ccccccc} 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 \end{array}
ight]$$

- Let $\mathbf{y} = (y_1 \ y_2 \dots \ y_n)$ be a received codeword.
- The syndrome of y is $S:=L_r y^T$. If S=0 then there was no error. If $S \neq 0$ then S is the binary representation of some integer $1 \le t \le n=2^r-1$ and the intended codeword is $\mathbf{x} = (y_1 \dots y_r+1 \dots y_n)$.



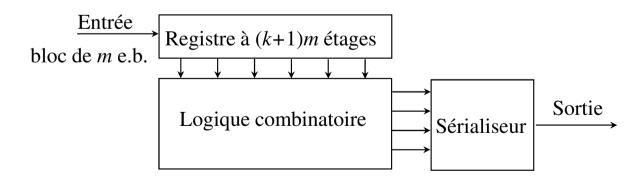


III- Convolutional Codes

- 1. Definition
- 2. Systematic Convolutional (SC) codes
- 3. Recursive Systematic Convolutional (RSC) Codes
- 4. Trellis termination

Convolutional codes

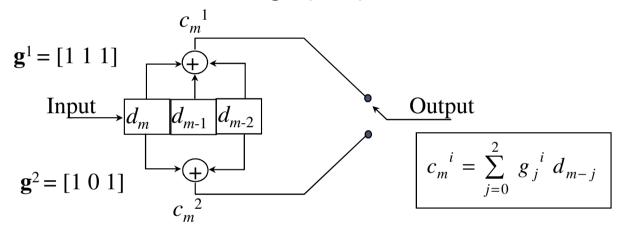
- A convolutional code is composed of
 - A register of (k+1).m stages storing (k+1) previous symbols,
 - a combinatorial logic calculating the blocks of *n bits* at the output.
- Constraint length: nombre of shifts needed in order to output an information block.
- Code rate: R = m / n.
- If m input bits are explicitly found in the block of n output bits, the code is then systematic.
- Convolutional coding is also used when the input bitstream is not separated into blocks after source coding.





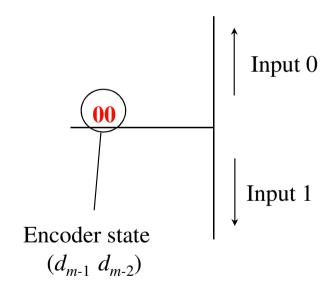
Example

• Code rate R = 1/2 and constraint length (k + 1) = 3.



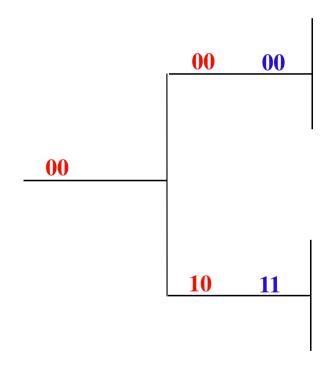
- The introduction of a memory effect complicates the representation of convolutional codes.
- We then adopt the representation in the form:
 - tree,
 - state diagram,
 - treillis.







$$\mathbf{E}$$
$$t = 0$$

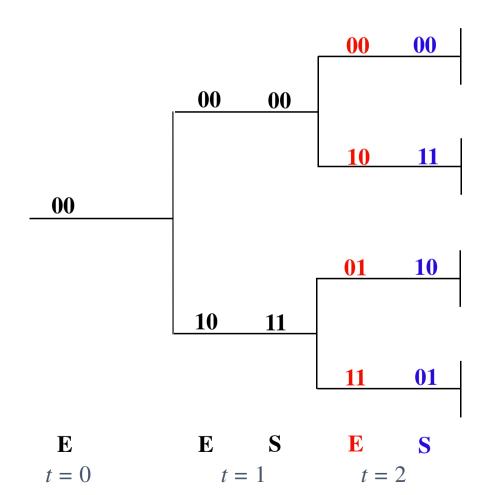




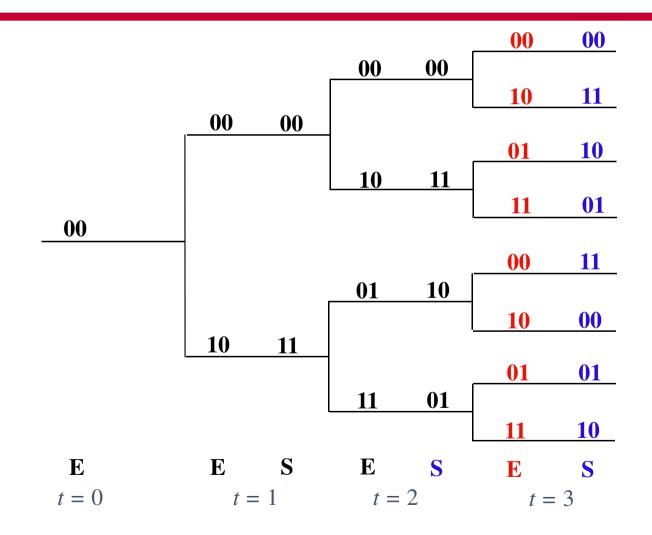
 \mathbf{E} t = 0

E

$$t = 1$$

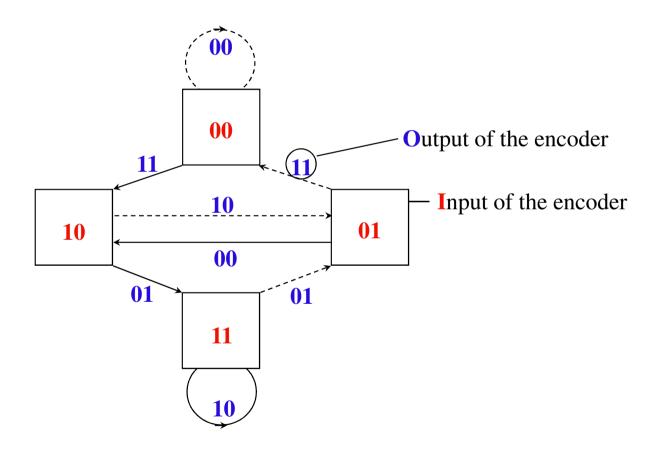




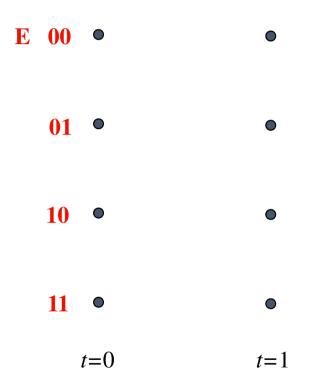




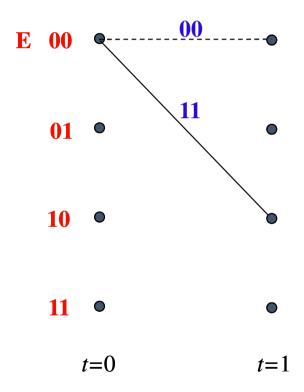
State diagram



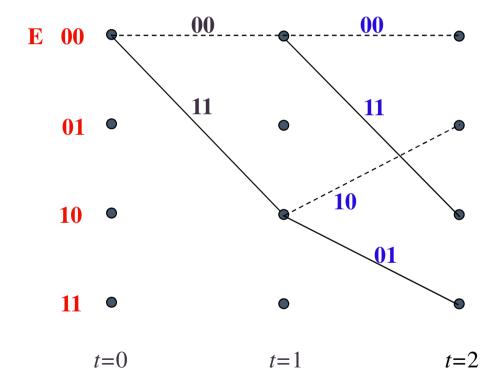




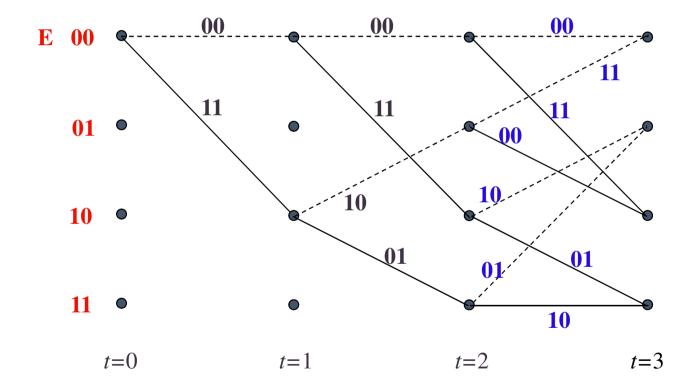




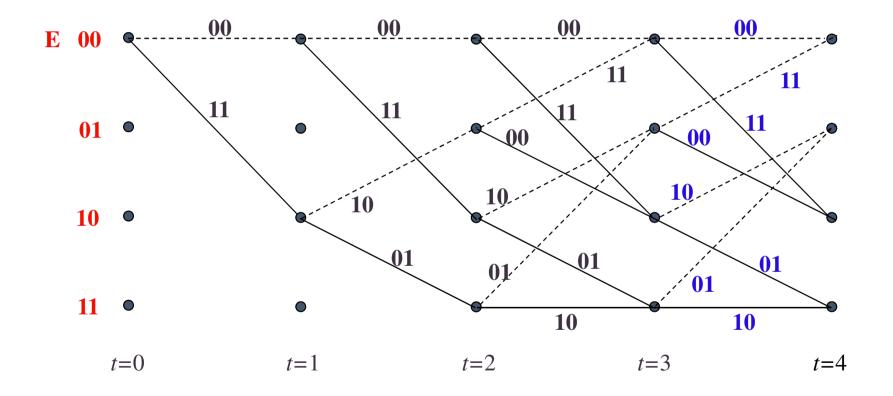






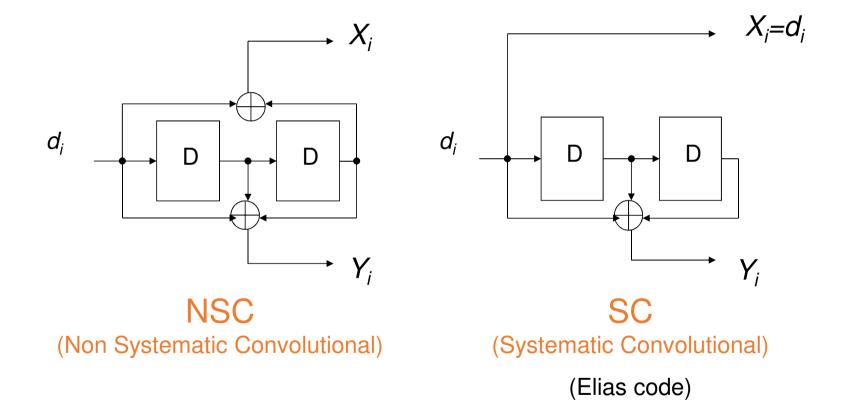






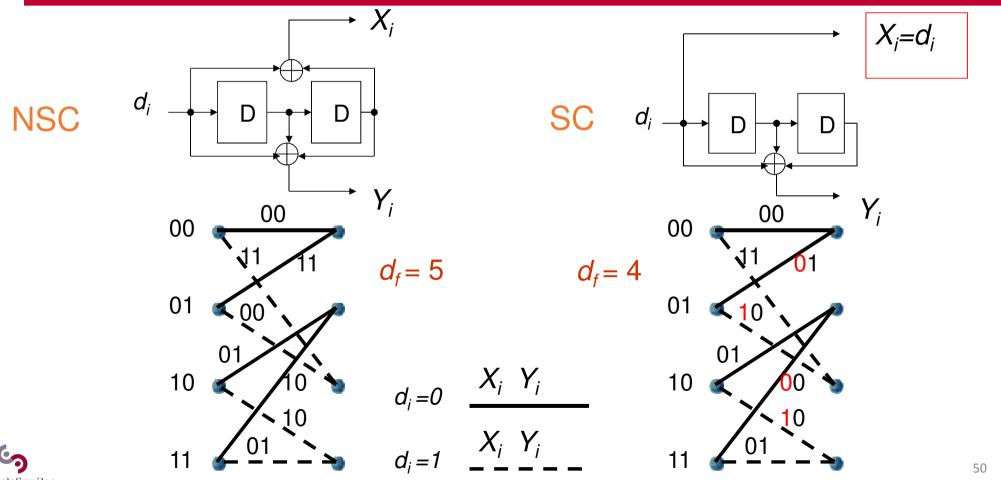


Systematic Convolutional (SC) codes





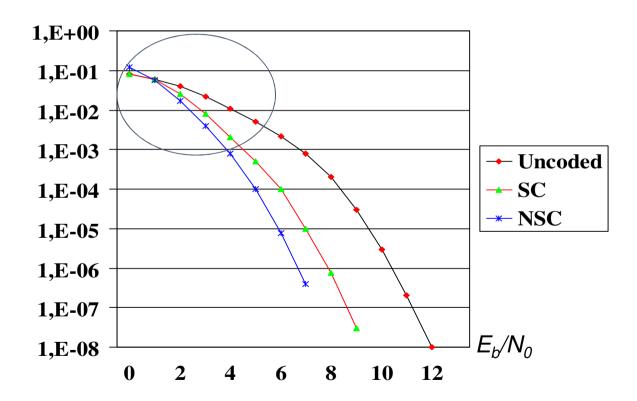
SC codes: trellis diagram





SC codes: Bit Error Rate

BER





SC codes: Weight Distribution

	n(d)	<i>w(d)</i> SC	n(d)	w(d) NSC
d=4	2	3	0	0
d=5	0	0	1	1
d=6	5	15	2	4
d=7	0	0	4	12
d=8	13	58	8	32
d=9	0	0	16	80
d=10	34	201	32	192
d=11	0		64	448
d=12	3 9	655	128	1024
d=13	0	0	256	2304
d=14	233	2052	512	5120

n(d) = number of cases

w(d)= total input weight

Asymptotic performance:

$$BER \le \frac{1}{2k} \sum_{d=d \text{ min}}^{\infty} w(d) \operatorname{erfc}(\sqrt{Rd \frac{E_b}{N_0}})$$
 (Union bound)

k = information sequence length R = coding rate E_b/N_0 = SNR

For high distances:

$$w(d)_{NSC} > w(d)_{SC}$$

=> the NSC code shows better performance for high noise levels



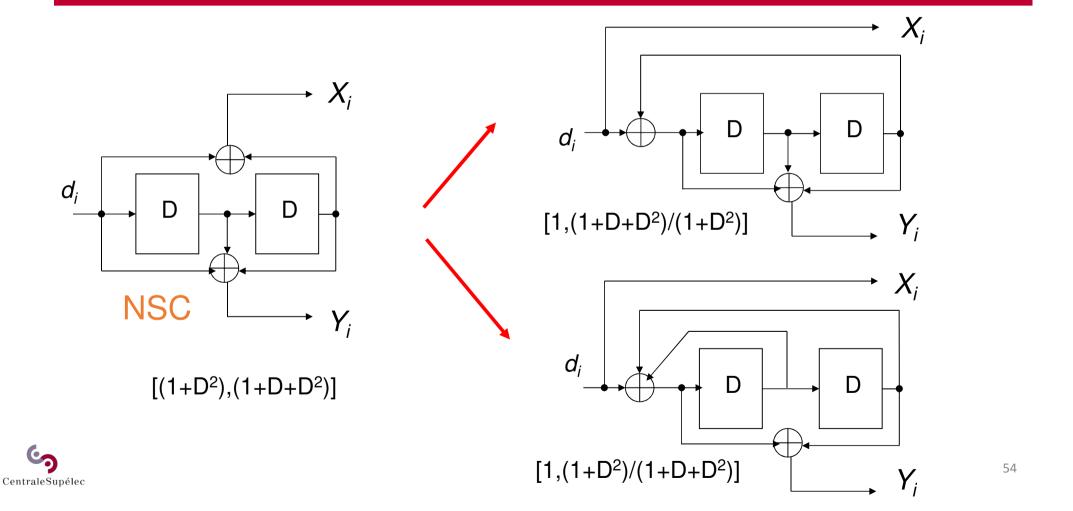
Is there any convolutional code able to combine

- the advantageous d_f of non systematic codes
- the good behaviour at very low SNRs

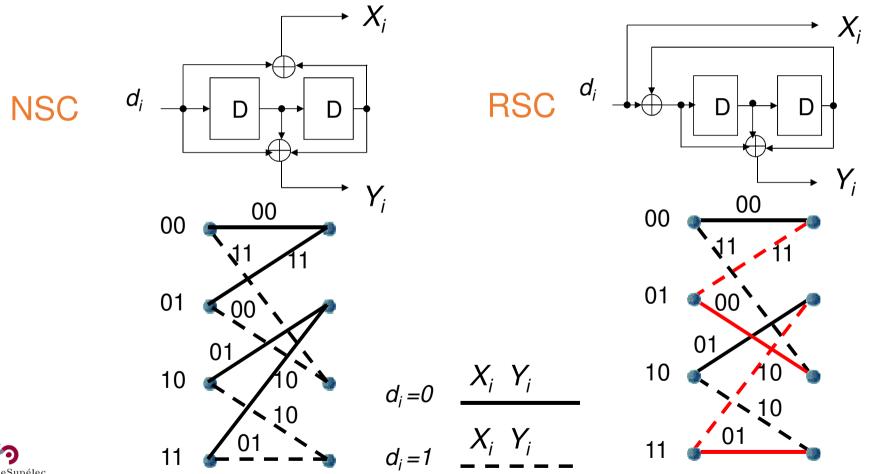
????

The answer is: yes

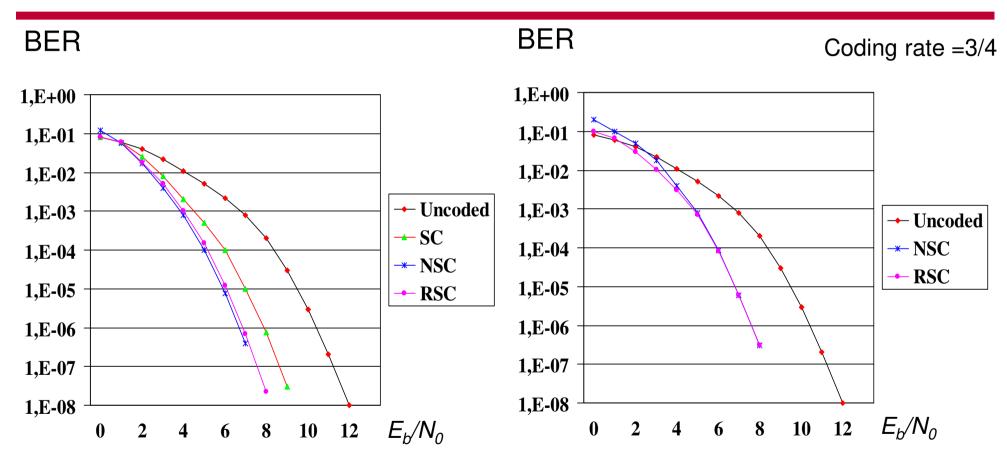
Recursive Systematic Convolutional (RSC) codes



RSC codes: Trellis diagram



RSC codes: Bit Error Rate



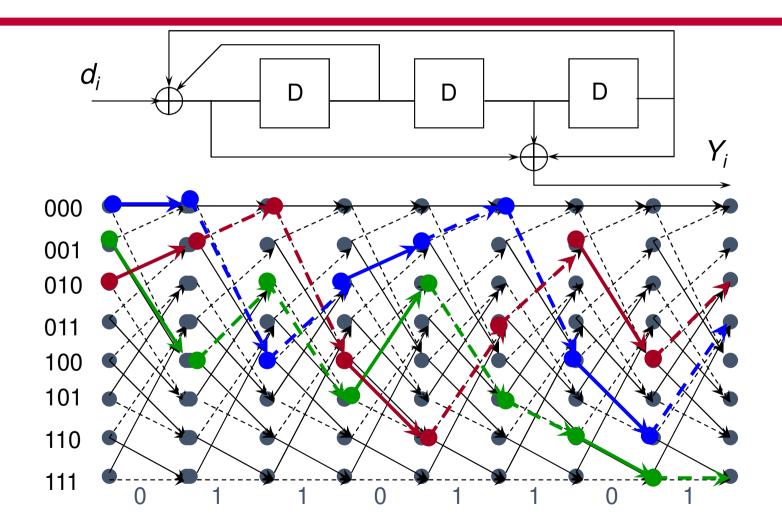


Circular Recursive Systematic Convolutional (CRSC) Codes: Trellis termination

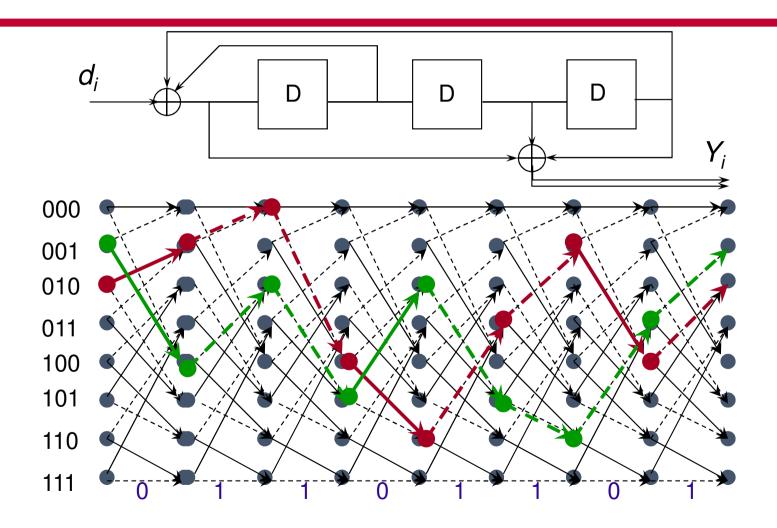
In many applications, block coding is required: How to transform a convolutional code into a block code?

- ➤ Inserting tail bits: encoder state returns to 0
 - easy to implement
 - the transmission of ν additional bits is needed
 - initial and final states are singular states
- Adopting circular encoding (= tail-biting)[9][10]
 - © coding rate remains unchanged
 - the trellis can be regarded as a circle without any singularity
 - a precoding step is necessary











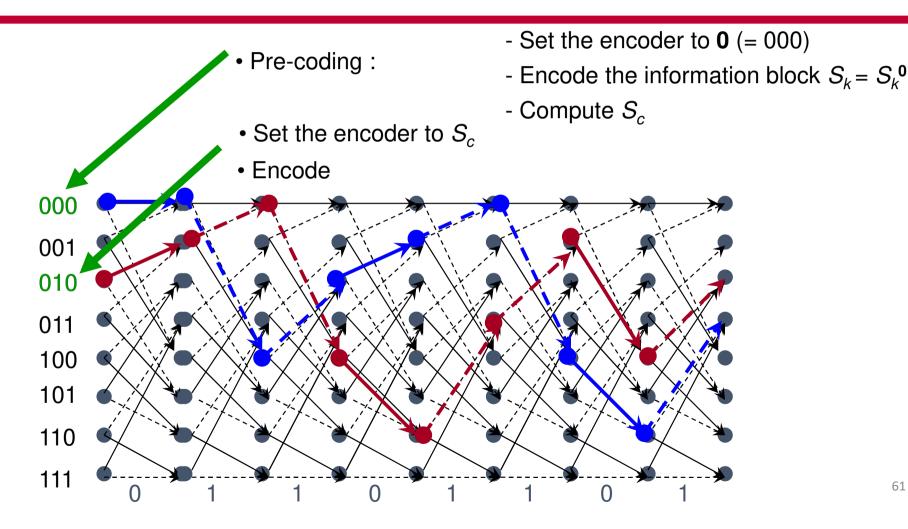
- It can be shown that, provided the data block length k is not a multiple of the encoder period P_{enc} ,
 - For a given information block, there is one and only one state S_c (circulation state) such as $S_c = S_0 = S_k$
 - S_c can be easily computed from

$$S_c = f(k \mod P_{enc}, S_k^0)$$

 S_k^0 is the final encoder state when encoding starts from state 0



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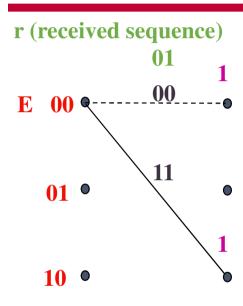
ML decoding of convolutional codes

- The decoding criterion for the convolutional codes is the same as for the codes in blocks: we will search the binary sequence c (corresponding to a path in the treillis) closest to the received sequence r, in the sense of the Hamming distance.
- This criterion is equivalent to maximum likelihood decoding
- Choose the more likely transmitted sequence

Viterbi algorithm

- Principle: search in the treillis diagram the path corresponding to the coded sequence with the minimum differences with received sequence
- The algorithm operates by computing a metric associated to all candidate paths
- The metric is simply the Hamming distance between the received sequence and the sequence relative to candidate path
- We retain as the most likely path the one whose metric is minimum.



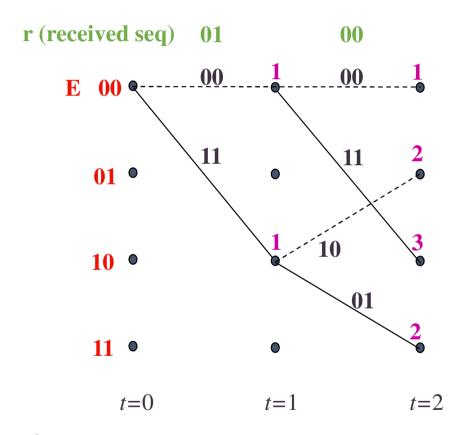




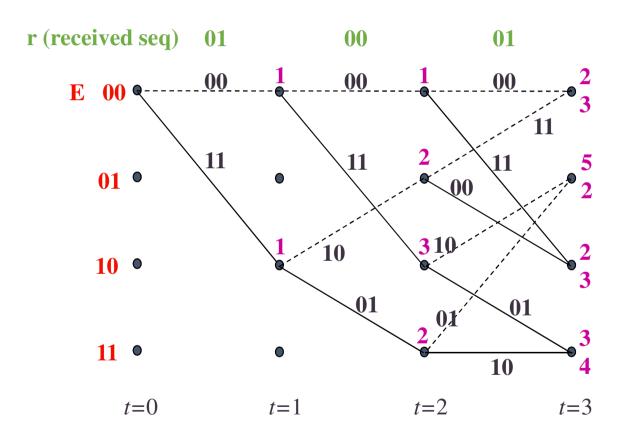
$$t=0$$
 $t=1$

- Example : Code de rendement R = 1/2 and constraint length (k + 1) = 3.
 - Transmitted sequence: **c** = (0000000000)
 - Received sequence: **r** = (0100010000)

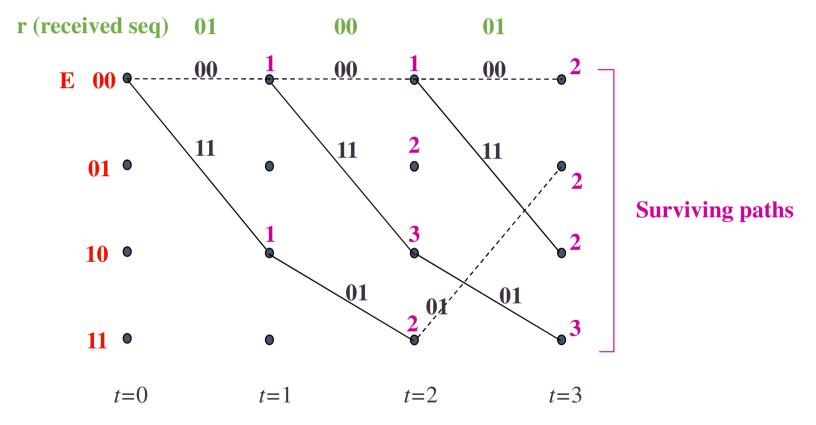




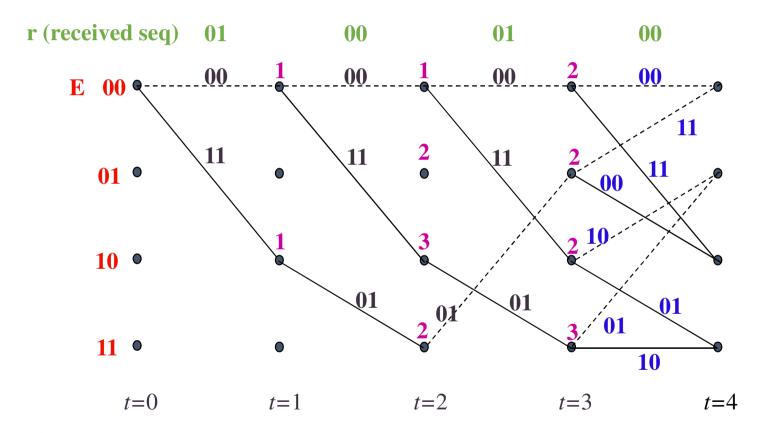




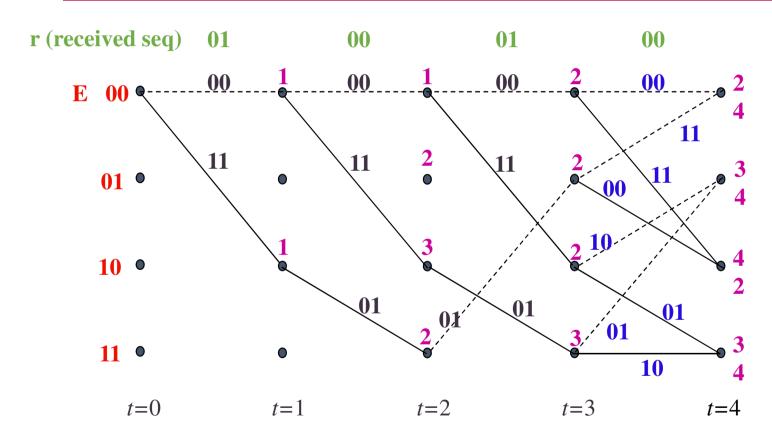




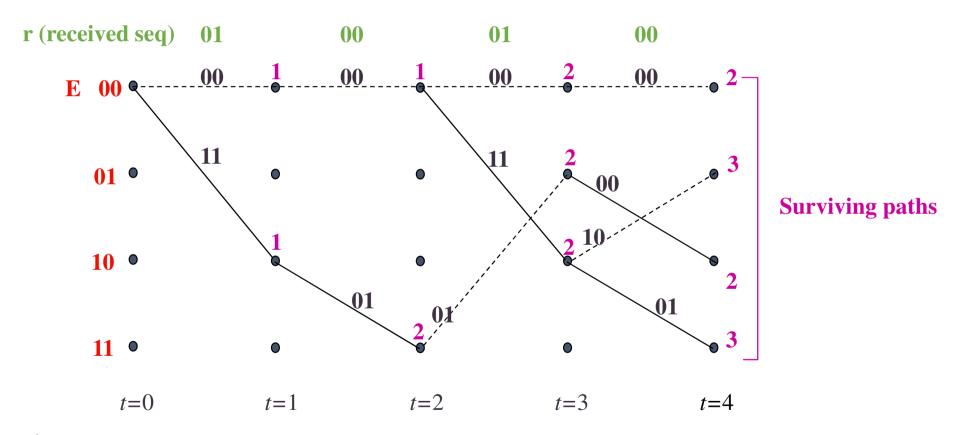




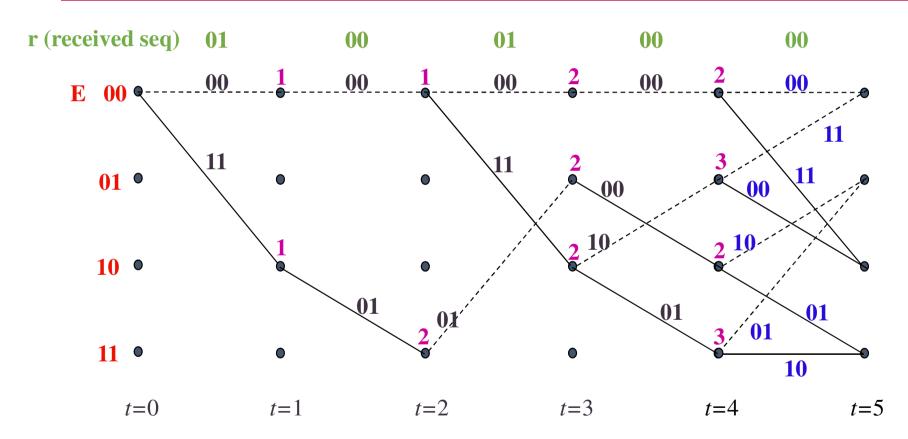




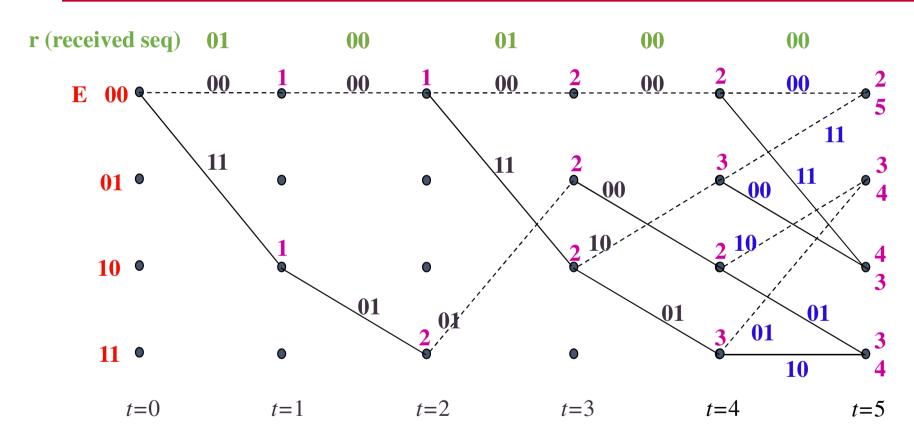




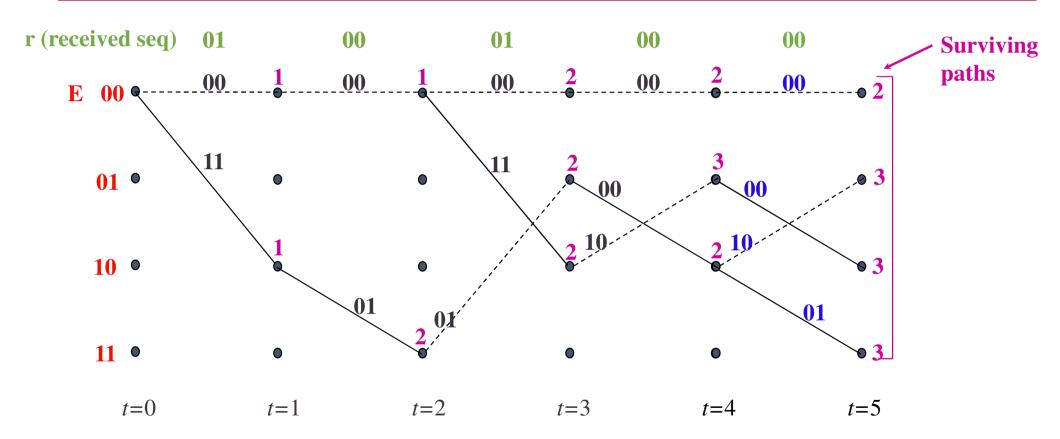






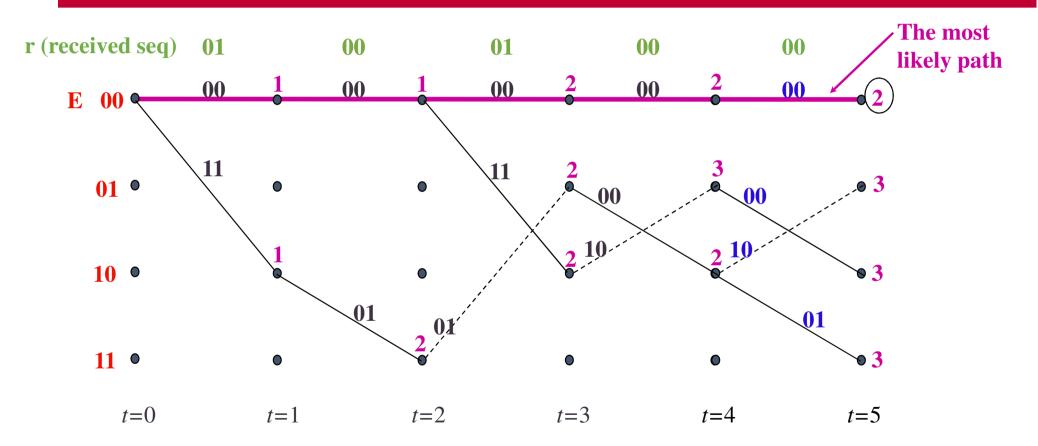








Viterbi decoding



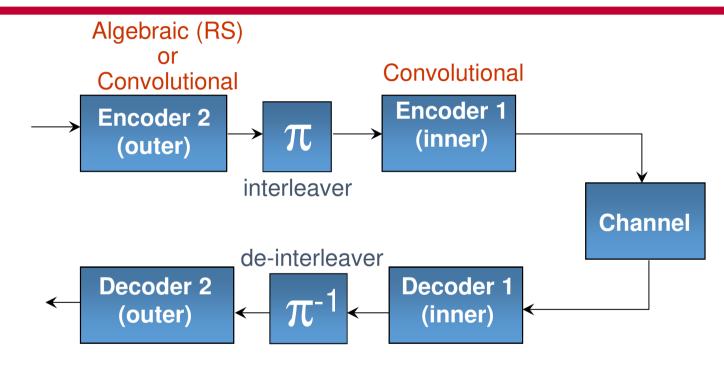




IV- Turbo-Codes

- 1. Concatenated codes
- 2. Turbo encoder
- 3. Turbo decoding
- 4. Peformance and standards

Concatenated Codes: divide and conquer



- First proposed by Forney in 1966 [1]
- Divide complexity between inner and outer decoding
- π and π^{-1} : de-interleaving breaks up error bursts at inner decoder output
- A standard in many communication systems: CCSDS, DVB-S1, DVB-T...



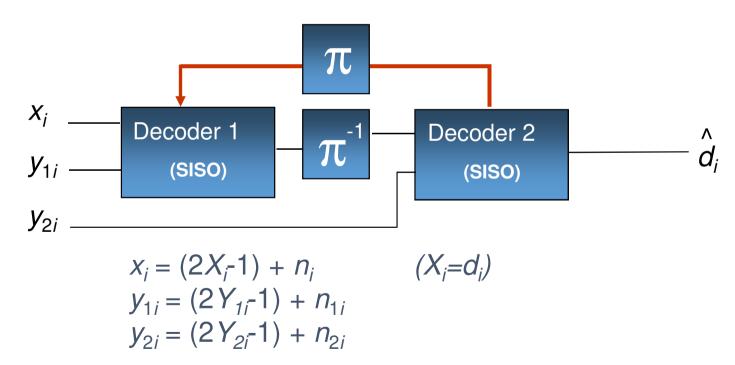
Concatenated Codes : performance improvements

- Performance of concatenated codes suffers because inner decoder only provides "hard" (0 or 1) decisions to outer decoder
- Step 1 Soft decision decoding (~1.5 to 2.0 dB gain)
 - Soft Output Viterbi Algorithm (SOVA): Battail (1987) [2] and Hagenauer (1989) [3]
 - Maximum a Posteriori (MAP) or BCJR algorithm (1974) [4]
- > Step2 "Turbo" decoding: Soft Iterative Decoding (<1 dB from channel capacity)
 - Turbo Codes (Parallel Concatenation of Convolutional Codes): Berrou and Glavieux (1993)
 - Turbo Product Codes: Pyndiah (1995)
 - Iterative decoding of Serially Concatenated Convolutional Codes: Benedetto *et al.* (1996) ...



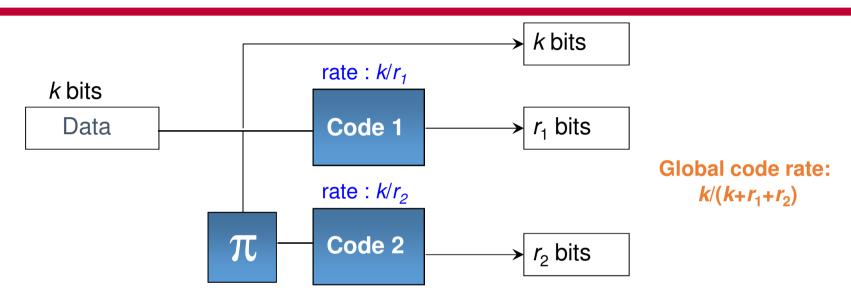
Iterative ("turbo") decoding of concatenated codes: principle

- Based on a Soft Input Soft Output (SISO) elementary decoder
- Iterative decoding process (turbo effect)





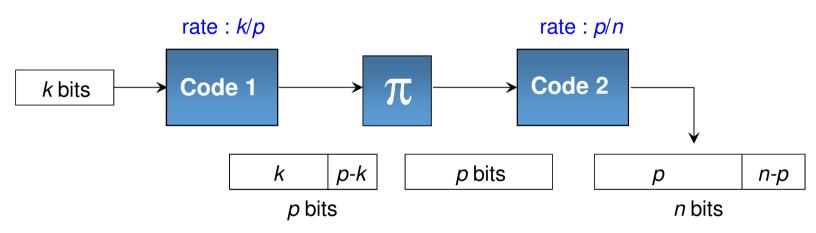
Parallel Concatenated Codes (PCC)



- Parallel concatenation of two (or more) systematic encoders separated by interleavers.
- Component codes can be algebraic or convolutional codes
- Historical Turbo Codes: code 1 and code 2 are two identical Recursive Systematic Convolutional (RSC) codes



Serial Concatenated Codes (SCC)

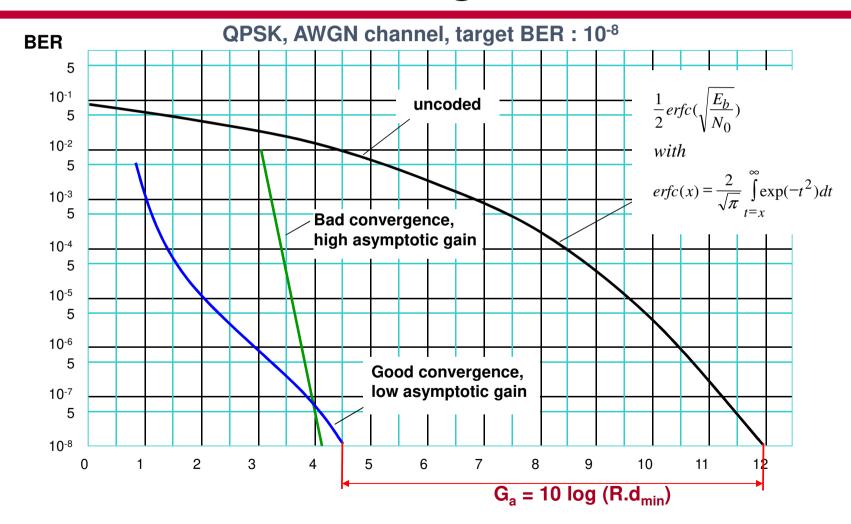


Global rate: k/n

- Component codes can be algebraic or convolutional codes
- SCCC: Serial Concatenation of Convolutional codes (Hagenauer)
- Some hybrid schemes (serial -//) can also be investigated (but have not found any practical application yet)

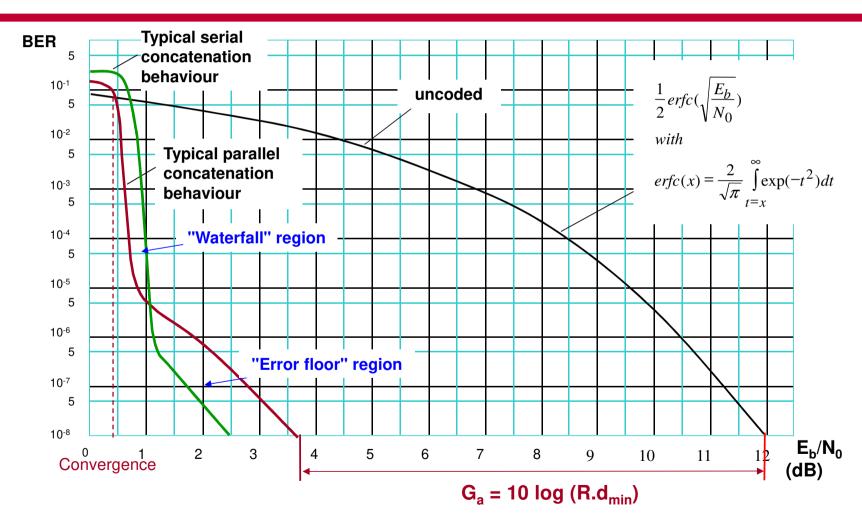


Qualitative behavior of concatenated codes under iterative decoding





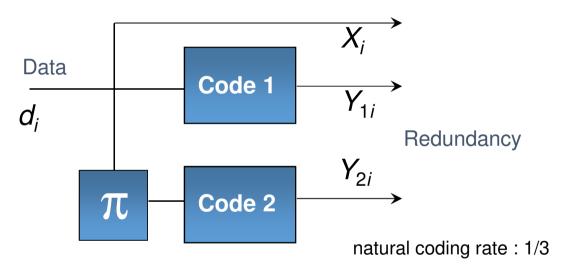
Parallel concatenation vs serial concatenation: performance comparison under iterative decoding





Turbo encoder (1)

The turbo encoder involves a parallel concatenation of at least two elementary Recursive Systematic Convolutional (RSC) codes separated by an interleaver (π)



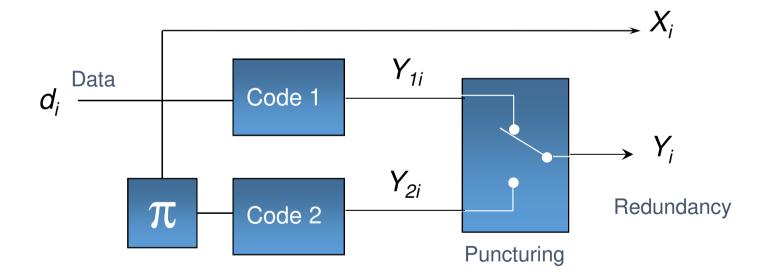
Parallel concatenation of two RSC codes or «convolutional» turbo code



Turbo encoder (2)

Puncturing for higher rates

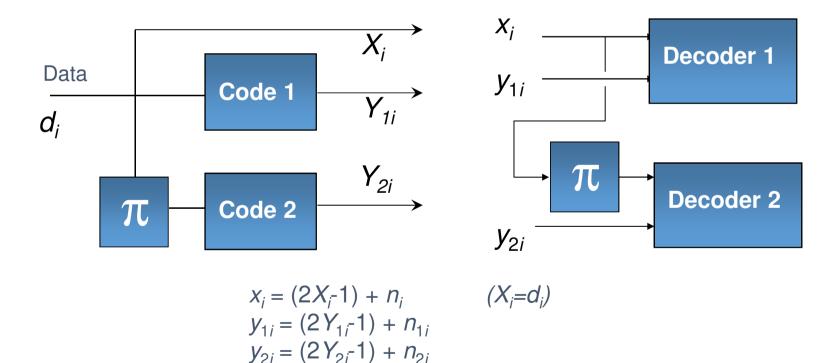
Example : R = 1/2





Turbo codes: Permutation

The basic idea: compensate for the vulnerability of a decoder to errors occuring in bursts





Turbo codes: Permutation

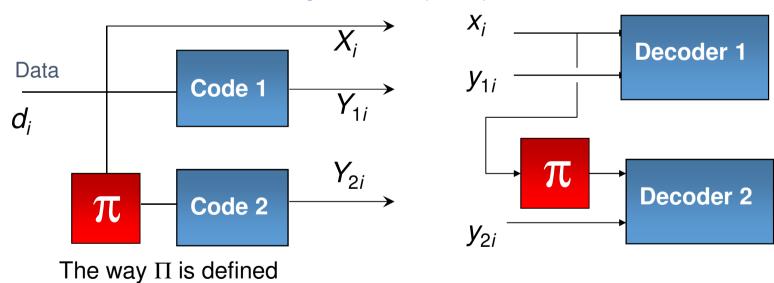
The fundamental role of π :

minimize the probability that both decoders fail (a very tricky problem combining algebra, geometry, ...)

=> Controls the Minimum Hamming Distance (MHD)

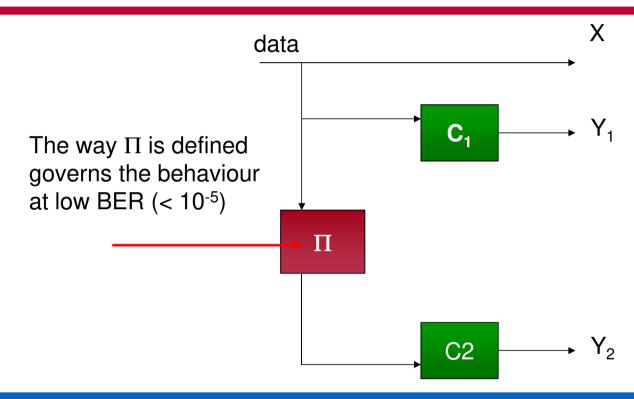
governs the behaviour

at low BER ($< 10^{-5}$)





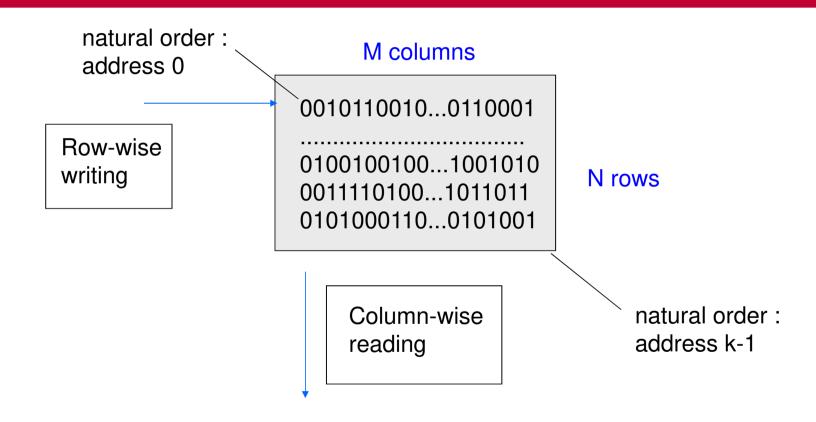
Permutation (permutation ≡ interleaving)



The fundamental role of the permutation: if the direct sequence is RTZ, minimize the probability that the permuted sequence is also RTZ and vice-versa.



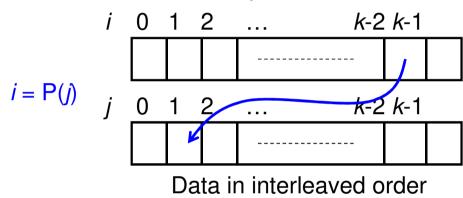
Regular permutation (1)





Regular permutation (2)

Another representation

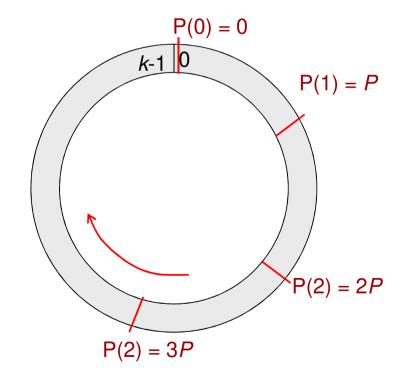


for
$$j = 0 \cdots k - 1$$

$$i = \Pi(j) = Pj \mod k$$

P and *k* are relatively prime integers

Data in natural order





Permutation

So, let us introduce disorder, but not in any manner!!

A good permutation must ensure:

- (1) maximum scattering of data
- (2) maximum disorder in the permuted sequence

(these two conditions are in conflict)



Permutation: Inserting controlled disorder

Everyone has his own tricks

For instance, the turbo code permutation DVB-RCS/RCT (ETSI EN 301 790 and 301 958), IEEE 802.16

♦ Built from a regular permutation

for
$$j = 0 \cdots N - 1$$

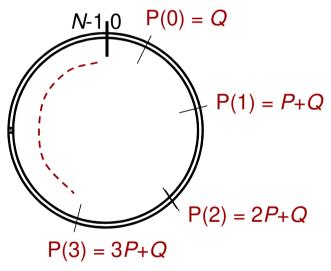
 $i = \Pi(j) = Pj + Q \mod N$

P and N are relatively prime integers

♦ Introduction of controlled disorder (here with max degree 4)

$$j \mod 4 = 0 \Rightarrow Q = P_1$$

 $j \mod 4 = 1 \Rightarrow Q = P_2$
 $j \mod 4 = 2 \Rightarrow Q = P_3$
 $j \mod 4 = 3 \Rightarrow Q = P_4$

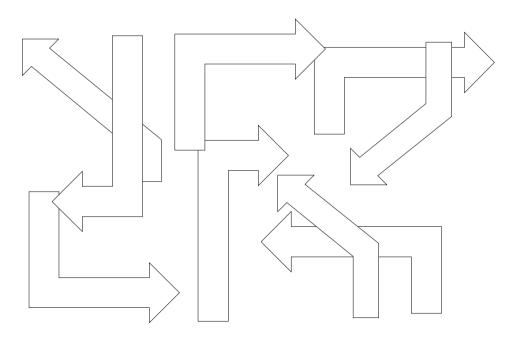


 $P_1 \dots P_4$ are small integers depending on the block size



Permutation

In conclusion:



No ideal permutation for the time being (does it exist?)



Horizontal

- I. Grecque
- II. Aéronef
- III. Réfutent
- IV. Chemin
- V. Anneaux

Vertical

- 1. Ballet
- 2. Lavabo
- 3. Attaches
- 4. Taille
- 5. Piliers

	1	2	3	4	5
ı	Т	Н	Е	Т	Α
ı II	Α	R	I	O	N
Ш	L	I	Е	S	Т
IV	S	Е	N	Т	N
V	Е	R	S	U	L

	1	2	3	4	5	_
1	Т	Н	Е	Т	Α	<u></u>
П	Α	R	I	0	N	8
Ш	L	I	Е	S	Т	8
IV	S	Е	N	Т	N	8
V	Е	R	S	U	L	8
'	8	8	8	8	(3)	8



Horizontal

- I. Grecque
- II. Aéronef
- III. Réfutent
- IV. Chemin
- V. Anneaux

Vertical

- 1. Ballet
- 2. Lavabo
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CentraleSupélec



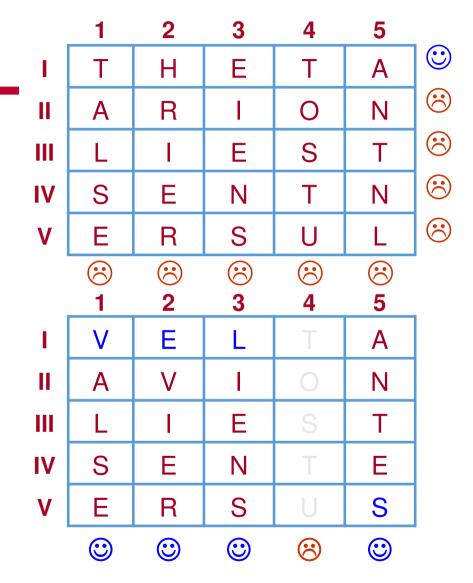


Horizontal

- I. Grecque
- II. Aéronef
- III. Réfutent
- IV. Chemin
- V. Anneaux

Vertical

- 1. Ballet
- 2. Lavabo
- 3. Attaches
- 4. Taille
- 5. Piliers





Vertical decoding

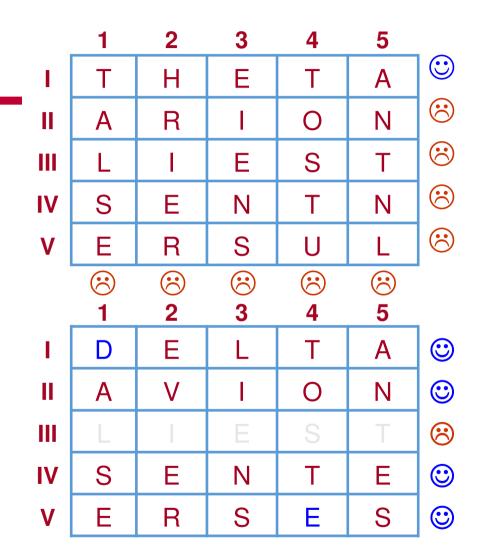
Horizontal

- I. Grecque
- II. Aéronef
- III. Réfutent
- IV. Chemin
- V. Anneaux

Vertical

- 1. Ballet
- 2. Lavabo
- 3. Attaches
- 4. Taille
- 5. Piliers

CentraleSupélec



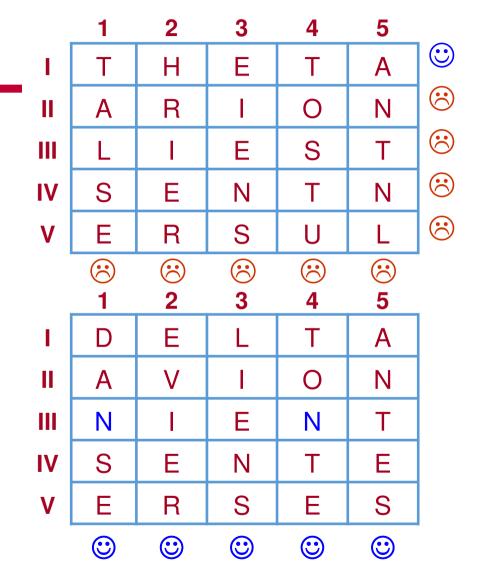


Horizontal

- I. Grecque
- II. Aéronef
- III. Réfutent
- IV. Chemin
- V. Anneaux

Vertical

- 1. Ballet
- 2. Lavabo
- 3. Attaches
- 4. Taille
- 5. Piliers





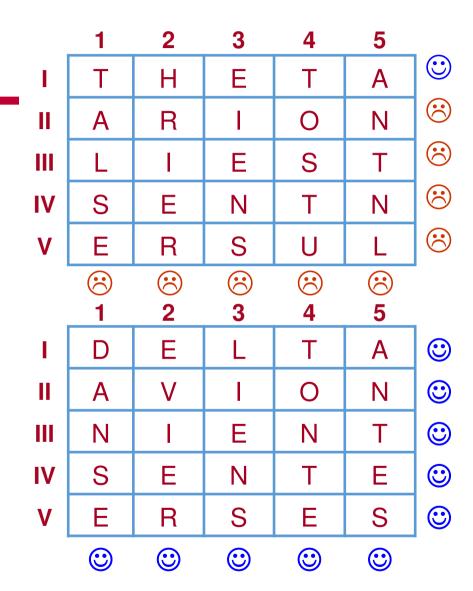
Vertical decoding

Horizontal

- I. Grecque
- II. Aéronef
- III. Réfutent
- IV. Chemin
- V. Anneaux

Vertical

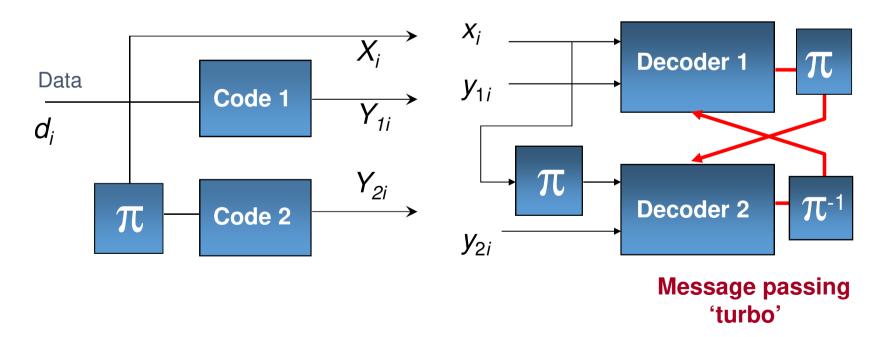
- 1. Ballet
- 2. Lavabo
- 3. Attaches
- 4. Taille
- 5. Piliers





Turbo decoding

How to make both decoders work jointly, so that decoder 1 could benefit from Y_2 and decoder 2 could benefit from Y_1 ?





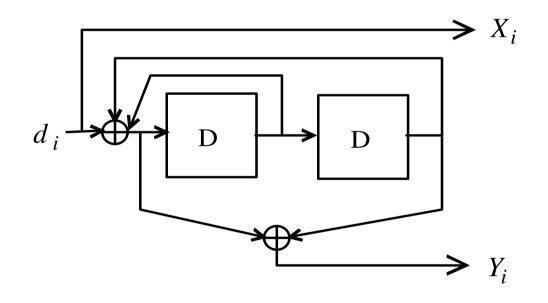
Soft-Input Soft-Output (SISO) decoding algorithms are necessary to turbo decoding.

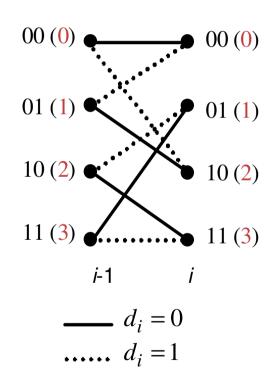
Two families of decoding algorithms:

- •The Viterbi based SISO algorithms [2][3]
- •The MAP algorithm and its approximations [4]



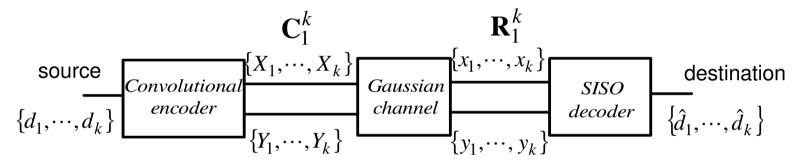
Illustration with the following code:







Considered transmission chain



The transmitted symbols X_i , $i=1\cdots k$ and Y_i , $i=1\cdots k$ are taken from set $\{-1,+1\}$

 \mathbb{C}_1^k is the encoded sequence

$$\mathbf{C}_{1}^{k} = \{C_{1}, \dots, C_{k}\} \text{ with } C_{i} = (X_{i}, Y_{i})$$

 \mathbf{R}_1^k is the noisy received sequence

$$\mathbf{R}_{1}^{k} = \{R_{1}, \dots, R_{k}\} \text{ with } R_{i} = (x_{i}, y_{i}), \ x_{i} = X_{i} + n_{i}^{x}, \ y_{i} = Y_{i} + n_{i}^{y}$$



The MAP (Maximum A Posteriori) algorithm

- Also known as BCJR (Bahl, Cocke, Jelinek, Raviv), APP (A Posteriori Probability), or Backward-Forward algorithm
- Aim of the algorithm: computation of the Logarithm of Likelihood Ratio (LLR) relative to data d_i

$$\Lambda(d_i) = \ln \frac{\Pr\{d_i = 1 \mid \mathbf{R}_1^k\}}{\Pr\{d_i = 0 \mid \mathbf{R}_1^k\}}$$

Sign ⇒ hard decision Magnitude ⇒ reliability

 \mathbf{R}_1^k is the noisy received sequence

$$\mathbf{R}_{1}^{k} = \{R_{1}, \dots, R_{k}\} \text{ with } R_{i} = (x_{i}, y_{i})$$



• Each *A Posteriori Probability* (APP) $\Pr\{d_i = j \mid \mathbf{R}_1^k\}$ can be written using joint probabilities $\lambda_i^j(\mathbf{m})$:

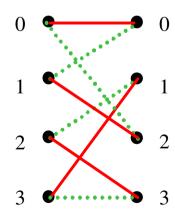
$$\lambda_i^j(\mathbf{m}) = \Pr\{d_i = j, \mathbf{S}_i = \mathbf{m}, \mathbf{R}_1^k\} \quad j = 0,1$$

 \mathbf{S}_i : encoder state at time or step i

$$\Pr\{d_i = j \, \middle| \, \mathbf{R}_1^k \} = \sum_{\mathbf{m}} \Pr\{d_i = j, \mathbf{S}_i = \mathbf{m} \, \middle| \, \mathbf{R}_1^k \} = \frac{\sum_{\mathbf{m}} \lambda_i^j(\mathbf{m})}{\Pr\{\mathbf{R}_1^k \}}$$



$$\Lambda(d_i) = \ln \frac{\lambda_i^1(\mathbf{0}) + \lambda_i^1(1) + \lambda_i^1(\mathbf{2}) + \lambda_i^1(\mathbf{3})}{\lambda_i^0(\mathbf{0}) + \lambda_i^0(\mathbf{1}) + \lambda_i^0(\mathbf{2}) + \lambda_i^0(\mathbf{3})}$$





- The principle of MAP algorithm: processing separately data available between steps 1 to i and between steps i+1 to k.
 - Introduction of *forward state probabilities* $\alpha_i(\mathbf{m})$, $i = 1 \cdots k$ => and backward state probabilities $\beta_i(\mathbf{m})$, $i = 1 \cdots k$

$$\alpha_i(\mathbf{m}) = \Pr{\{\mathbf{S}_i = \mathbf{m}, \mathbf{R}_1^i\}}$$

$$\alpha_i(\mathbf{m}) = \Pr{\{\mathbf{S}_i = \mathbf{m}, \mathbf{R}_1^i\}}$$

$$\beta_i(\mathbf{m}) = \Pr{\{\mathbf{R}_{i+1}^k | \mathbf{S}_i = \mathbf{m}\}}$$



One can show that:

$$\lambda_i^j(\mathbf{m}) = \sum_{\mathbf{m'}} \gamma_j(R_i, \mathbf{m'}, \mathbf{m}) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m}), j = 0,1$$

where $\gamma_j(R_i, \mathbf{m'}, \mathbf{m})$ represents the *branch probability* between states $\mathbf{m'}$ and \mathbf{m} when the received symbol is R_i at step i.

$$=> \frac{\sum \sum \gamma_1(R_i, \mathbf{m'}, \mathbf{m}) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})}{\sum \sum \gamma_0(R_i, \mathbf{m'}, \mathbf{m}) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})}$$

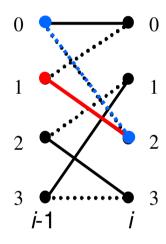


Example:

*
$$\lambda_i^0(\mathbf{2}) = \gamma_0(R_i, \mathbf{1}, \mathbf{2})\alpha_{i-1}(\mathbf{1})\beta_i(\mathbf{2})$$

*
$$\lambda_i^1(\mathbf{2}) = \gamma_1(R_i, \mathbf{0}, \mathbf{2}) \alpha_{i-1}(\mathbf{0}) \beta_i(\mathbf{2})$$

$$\Lambda(d_{i}) = \ln \frac{\gamma_{1}(R_{i}, \mathbf{1}, \mathbf{0})\alpha_{i-1}(\mathbf{1})\beta_{i}(\mathbf{0}) + \gamma_{1}(R_{i}, \mathbf{2}, \mathbf{1})\alpha_{i-1}(\mathbf{2})\beta_{i}(\mathbf{1})}{\gamma_{0}(R_{i}, \mathbf{0}, \mathbf{0})\alpha_{i-1}(\mathbf{0})\beta_{i}(\mathbf{0}) + \gamma_{0}(R_{i}, \mathbf{3}, \mathbf{1})\alpha_{i-1}(\mathbf{3})\beta_{i}(\mathbf{1})} \cdots
\cdots \frac{+\gamma_{1}(R_{i}, \mathbf{0}, \mathbf{2})\alpha_{i-1}(\mathbf{0})\beta_{i}(\mathbf{2}) + \gamma_{1}(R_{i}, \mathbf{3}, \mathbf{3})\alpha_{i-1}(\mathbf{3})\beta_{i}(\mathbf{3})}{+\gamma_{0}(R_{i}, \mathbf{1}, \mathbf{2})\alpha_{i-1}(\mathbf{1})\beta_{i}(\mathbf{2}) + \gamma_{0}(R_{i}, \mathbf{2}, \mathbf{3})\alpha_{i-1}(\mathbf{2})\beta_{i}(\mathbf{3})}$$



$$d_i = 0$$
 $d_i = 1$



• Forward recursion: computation of $\alpha_i(\mathbf{m})$

 $\alpha_i(\mathbf{m})$ can be recursively computed from

$$\alpha_i(\mathbf{m}) = \sum_{\mathbf{m'}} \sum_j \alpha_{i-1}(\mathbf{m'}) \gamma_j(R_i, \mathbf{m'}, \mathbf{m})$$

 $\alpha_i(\mathbf{m})$ are all computed during the forward recursion in the trellis from initial values $\alpha_0(\mathbf{m})$

If \mathbf{m}_0 is the initial state of the encoder: $\alpha_0(\mathbf{m}_0) = 1$ $\alpha_0(\mathbf{m}) = 0 \quad \forall \mathbf{m} \neq \mathbf{m}_0$

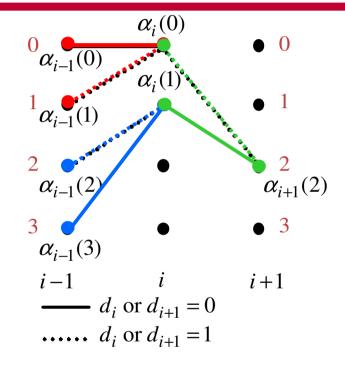
NB: to avoid numerical precision problems, $\alpha_i(\mathbf{m})$ can be normalized at each step of processing $(\Lambda(d_i))$ is a ratio

$$\alpha_i'(\mathbf{m}) = \frac{\alpha_i(\mathbf{m})}{\sum_{\mathbf{m}'} \alpha_i(\mathbf{m}')}$$



Example:

(without normalization)



*
$$\alpha_i(\mathbf{0}) = \alpha_{i-1}(\mathbf{0})\gamma_0(R_i, \mathbf{0}, \mathbf{0}) + \alpha_{i-1}(\mathbf{1})\gamma_1(R_i, \mathbf{1}, \mathbf{0})$$

*
$$\alpha_i(1) = \alpha_{i-1}(3)\gamma_0(R_i, 3, 1) + \alpha_{i-1}(2)\gamma_1(R_i, 2, 1)$$

*
$$\alpha_{i+1}(\mathbf{2}) = \alpha_i(\mathbf{1})\gamma_0(R_{i+1},\mathbf{1},\mathbf{2}) + \alpha_i(\mathbf{0})\gamma_1(R_{i+1},\mathbf{0},\mathbf{2})$$

Backward recursion: computation of $\beta_i(\mathbf{m'})$

$$\beta_i(\mathbf{m'})$$
 can be recursively computed from
$$\beta_i(\mathbf{m'}) = \sum_{\mathbf{m'}} \sum_j \beta_{i+1}(\mathbf{m}) \gamma_j(R_{i+1}, \mathbf{m'}, \mathbf{m})$$

 β_i (m') are all computed during the backward recursion into the trellis from initial values $\beta_k(\mathbf{m})$

If the final state of the encoder is known (\mathbf{m}_k):

$$\beta_k(\mathbf{m}_k) = 1, \quad \beta_k(\mathbf{m}) = 0 \quad \forall \mathbf{m} \neq \mathbf{m}_k$$

If the final state of the encoder is unknown:

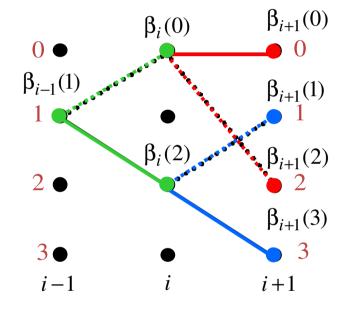
$$\beta_k(\mathbf{m}) = \frac{1}{2^{\nu}} \quad \forall \mathbf{m} \quad \text{(n = code memory)}$$

Normalization:
$$\beta_i'(\mathbf{m}) = \frac{\beta_i(\mathbf{m})}{\sum_{\mathbf{m}'} \beta_i(\mathbf{m}')}$$



Example:

(without normalization)



$$d_i \text{ or } d_{i+1} = 0$$

$$d_i \text{ or } d_{i+1} = 1$$

*
$$\beta_i(\mathbf{0}) = \beta_{i+1}(\mathbf{0})\gamma_0(R_{i+1},\mathbf{0},\mathbf{0}) + \beta_{i+1}(\mathbf{2})\gamma_1(R_{i+1},\mathbf{0},\mathbf{2})$$

*
$$\beta_i(\mathbf{2}) = \beta_{i+1}(\mathbf{3})\gamma_0(R_{i+1},\mathbf{2},\mathbf{3}) + \beta_{i+1}(\mathbf{1})\gamma_1(R_{i+1},\mathbf{2},\mathbf{1})$$

*
$$\beta_{i-1}(\mathbf{1}) = \beta_i(\mathbf{2})\gamma_0(R_i,\mathbf{1},\mathbf{2}) + \beta_i(\mathbf{0})\gamma_1(R_i,\mathbf{1},\mathbf{0})$$



• Branch probabilities: computation of $\gamma_j(R_i, \mathbf{m'}, \mathbf{m})$

If there is no transition between \mathbf{m}' and \mathbf{m} in the trellis or if the transition is not labeled with $d_i = j$.

$$\Rightarrow \gamma_j(R_i, \mathbf{m'}, \mathbf{m}) = 0$$

else, for a transmission over a Gaussian channel

$$\gamma_j(R_i, \mathbf{m'}, \mathbf{m}) = \Pr\{d_i = j\} \times \frac{1}{2\pi\sigma^2} \exp\left(-\frac{\|R_i - C_i\|^2}{2\sigma^2}\right)$$

Where C_i is the expected symbol along the branch from state \mathbf{m}' to state \mathbf{m} and R_i is the received symbol .



• Branch probabilities: computation of $\gamma_j(R_i, \mathbf{m'}, \mathbf{m})$

If
$$C_i = (X_i, Y_i)$$
, $R_i = (x_i, y_i)$ and $\Pr\{d_i = 0\} = \Pr\{d_i = 1\} = \frac{1}{2}$

$$\gamma_j(R_i, \mathbf{m'}, \mathbf{m}) = \frac{1}{4\pi\sigma^2} \exp\left(-\frac{(x_i - X_i)^2}{2\sigma^2}\right) \exp\left(-\frac{(y_i - Y_i)^2}{2\sigma^2}\right)$$

$$\gamma_{1}(R_{i}, \mathbf{m'}, \mathbf{m}) = A \exp\left(\frac{x_{i}}{\sigma^{2}}\right) \exp\left(\pm \frac{y_{i}}{\sigma^{2}}\right) \quad (X_{i} = +1)$$

$$\gamma_{0}(R_{i}, \mathbf{m'}, \mathbf{m}) = A \exp\left(-\frac{x_{i}}{\sigma^{2}}\right) \exp\left(\mp \frac{y_{i}}{\sigma^{2}}\right) \quad (X_{i} = -1)$$



Derivation of extrinsic information for iterative decoding

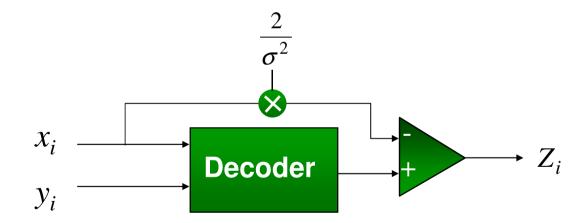
$$\begin{split} & \Lambda(d_i) = \ln \frac{\sum \sum \gamma_1(R_i, \mathbf{m'}, \mathbf{m}) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})}{\sum \sum \sum \gamma_0(R_i, \mathbf{m'}, \mathbf{m}) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})} \\ & \Lambda(d_i) = \ln \frac{\sum \sum \sum \exp \left(\frac{x_i}{\sigma^2}\right) \exp \left(\pm \frac{y_i}{\sigma^2}\right) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})}{\sum \sum \sum \exp \left(-\frac{x_i}{\sigma^2}\right) \exp \left(\mp \frac{y_i}{\sigma^2}\right) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})} \end{split}$$

$$\Lambda(d_i) = \frac{2}{\sigma^2} x_i + \ln \frac{\sum_{\mathbf{m}} \sum_{\mathbf{m'}} \exp\left(\pm \frac{y_i}{\sigma^2}\right) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})}{\sum_{\mathbf{m}} \sum_{\mathbf{m'}} \exp\left(\mp \frac{y_i}{\sigma^2}\right) \alpha_{i-1}(\mathbf{m'}) \beta_i(\mathbf{m})}$$



Derivation of extrinsic information for iterative decoding

$$\Lambda'(d_i)$$
 can be written as $\Lambda'(d_i) = \frac{2}{\sigma^2}x_i + Z_i$

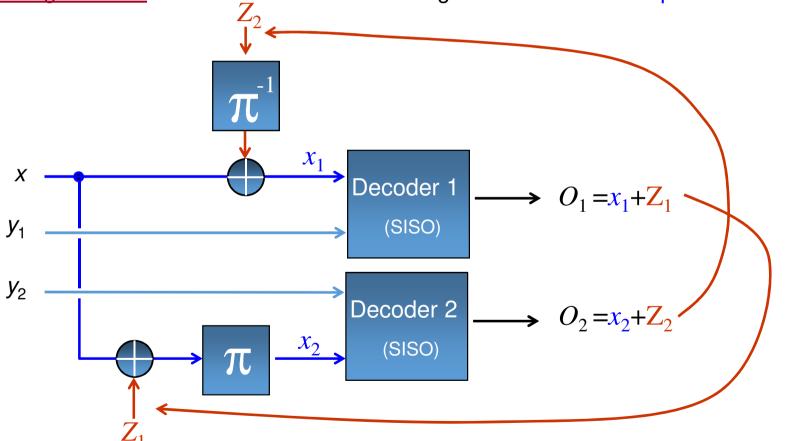


 Z_i is the *extrinsic information* provided by the decoder



TC: Extrinsic Information

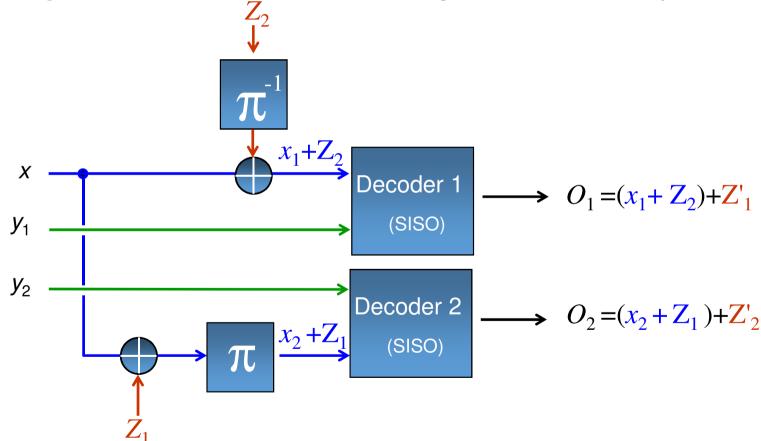
Decoding criterion: both decoders have to merge towards the same probabilistic decision





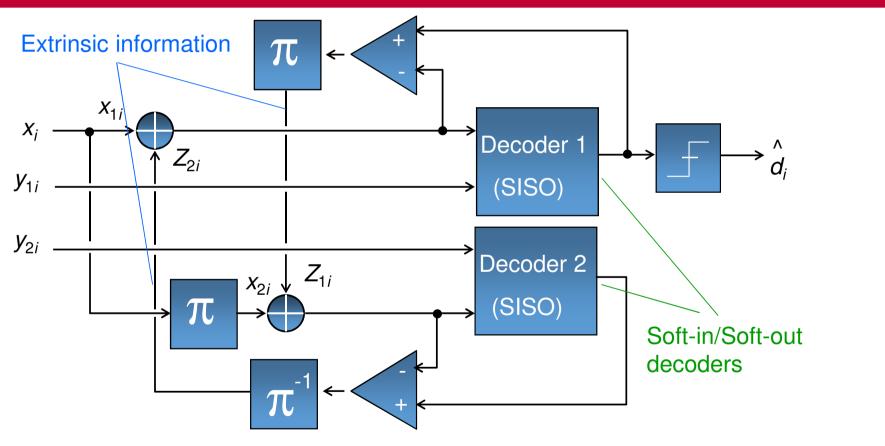
TC: Extrinsic Information

• Decoding criterion: both decoders have to merge towards the same probabilistic decision





The Turbo decoder structure



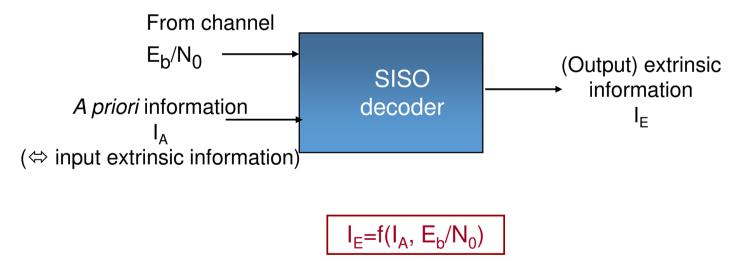


Through extrinsic information, each decoder uses both redundancies.

Fundamental principle: a decoder must not reuse a piece of information provided by itself.

Iterations and convergence of a turbo decoder

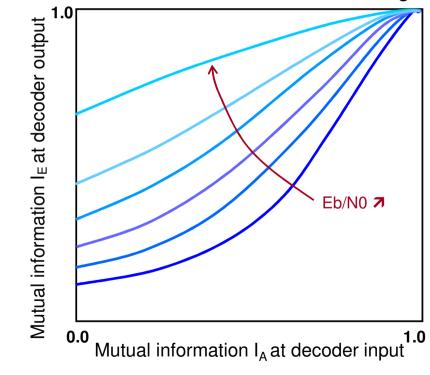
- Mutual information transfer characteristic
 - Describes the flow of extrinsic information through the SISO decoder





Iterations and convergence of a turbo decoder

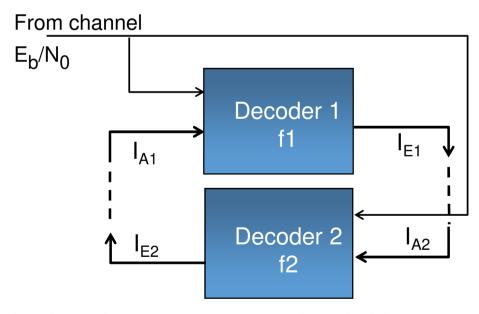
- Mutual information transfer characteristic
 - Describes the flow of extrinsic information through the SISO decoder





Iterations and convergence of a turbo decoder

- Mutual information transfer characteristic
 - Describes the exchange of extrinsic information between the SISO decoders

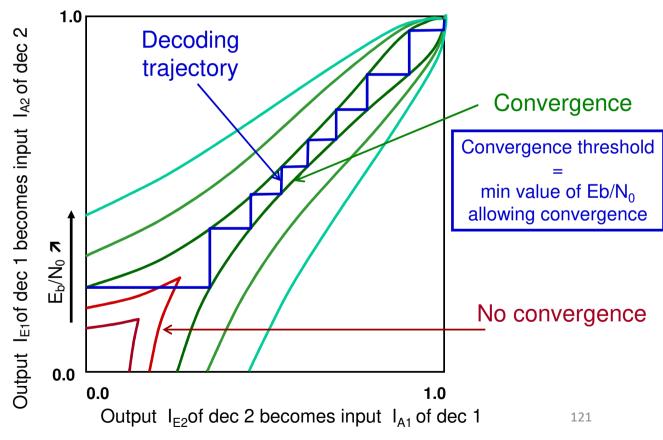


Determination of the convergence threshold



Iterations and convergence of a turbo decoder

- Mutual information transfer characteristic
 - Describes the exchange of extrinsic information between the SISO decoders





TC: Examples of performance

Received image





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TC: Examples of performance After 1 decoding iteration



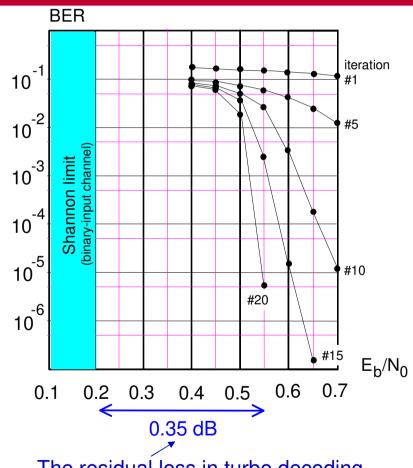


TC: Examples of performance After 8 decoding iteration





Convolutional Turbo Codes



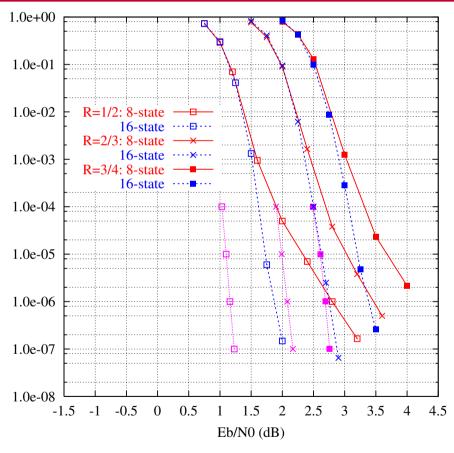
Example of performance

- QPSK modulation
- R=1/2
- AWGN channel
- Double-binary 8-state turbo code
- Large block size
- MAP algorithm
- No quantization



The residual loss in turbo decoding

Convolutional Turbo Codes



Example of performance

- QPSK modulation
- AWGN channel
- Double-binary turbo codes
- 1504-bit data blocks
- Max-log-MAP algorithm
- 4-bit quantization
- 8 iterations



Turbo Codes: applications





Turbo Codes: standards (1/2)

Application	Turbo code	termination	polynomials	rates
CCSDS (deep space)	binary, 16- state	tail bits	23,33,25,37	1/6,1/4,1/3,1/2
3GPP (UMTS)	binary, 8-state	tail bits	13,15,17	1/4,1/3,1/2
3GPP2 (CDMA2000)	binary, 8-state	tail bits	13,15,17	1/4,1/3,1/2
3GPP LTE (Long Term Evolution)	binary, 8-state	tail bits	13,15,17	1/4,1/3,1/2
DVB-RCS (Return Channel over Satellite)	double-binary, 8-state	circular (tail- biting)	15,13	1/3 up to 6/7
DVB-RCT (Return Channel over Terrestrial)	double-binary, 8-state	circular (tail- biting)	15,13	1/2, 3/4
DVB-SSP (satellite service to portable)	binary, 8-state	tail bits	15,13	



Turbo Codes: standards (2/2)

Application	Turbo code	termination	polynomials	rates
IEEE 802.16 (WiMAX) + IEEE 802.16e (Mobile WiMAX)	double-binary, 8-state +Turbo product code	circular (tail-biting)	15,13	
Inmarsat (Aero-H)	binary, 16-state	No	23, 35	
Eutelsat (Skyplex)	double-binary, 8-state	circular (tail-biting)	15,13	4/5,6/7
Broadcom (Echostar)				
Qualcomm (MediaFLO)				
Homeplug	Turbo product code			





V- Error Detection and retransmission protocols

- 1. Cyclic Redundancy Check Codes
- 2. HARQ protocols
- 3. IR-HARQ protocols

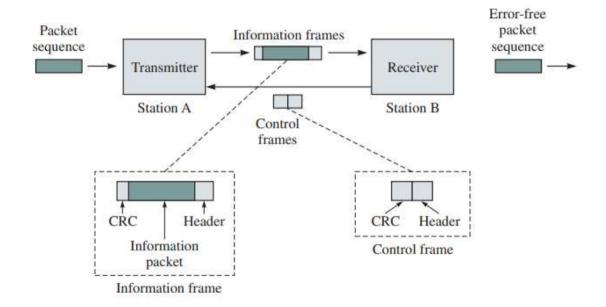
CRC (Cyclic Redundancy Check Codes)

- Ils peuvent détecter :
 - toutes les salves d'erreur de longueur *r* ou moins.
 - une partie des salves d'erreurs de longueur r + 1
 - une partie des salves d'erreurs de longueur > r + 1
 - toutes les combinaisons d'erreurs de poids d_{min}-1
 - toutes les combinaisons d'erreurs de poids impair si g(z) correspond à un motcode de poids pair.
- Les CRC possèdent tous (1+z) comme polynôme diviseur.



Retransmission protocols (1/2)

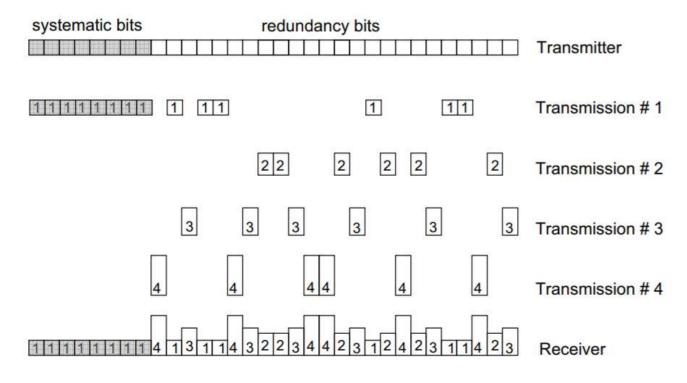
■ ARQ (Automatic Repeat reQuest) protocols





Retransmission protocols (2/2)

- Hybrid FEC/ARQ protocols
 - Type I HARQ (same channel coding + chase combining)
 - Type II HARQ (Incremental redundancy)







Thank you for you attention!

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