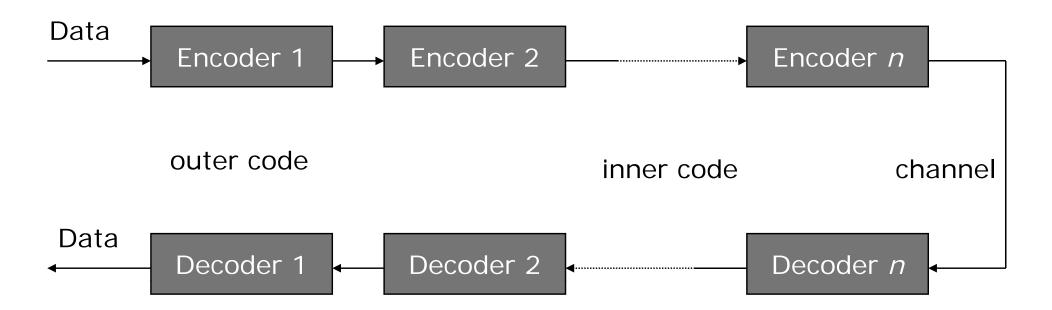
# Principle of Turbo Codes

#### Concatenated Codes

- → The power of forward error correcting (FEC) codes increases with length k.
- Decoding complexity also increases very rapidly with length k.
- Solve the problem by building a long, complex code out of much shorter component codes, which can be decoded much more easily.
- → This is called concatenation technique.

#### Principle of Concatenated Codes

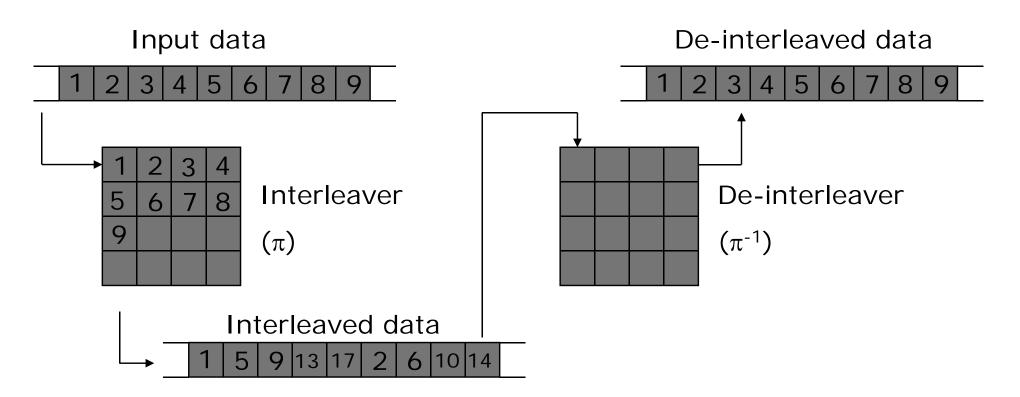
Serial concatenation



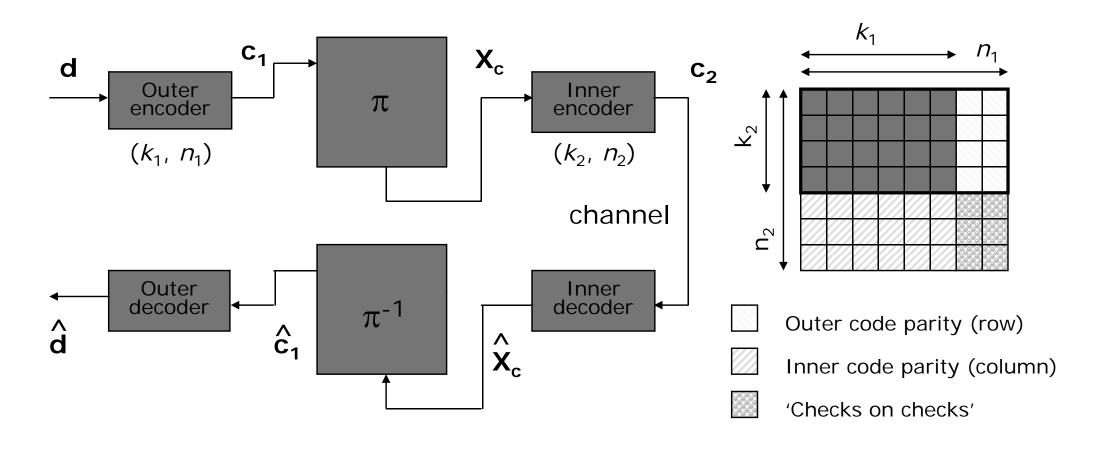
- Most significant drawback: error propagation
- Decoding error in a codeword of one decoder will be passed to the next decoder
- Too many errors in one codeword may overwhelm the decoder to correct the error

- → Improve performance by distributing these errors to a number of separate codewords before inputting into the next decoder
- This is achieved using interleaver and de-interleaver

Block or rectangular interleaver

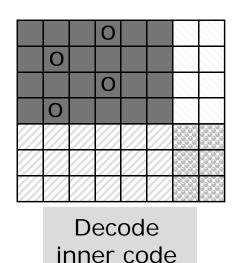


Concatenated code with interleaver



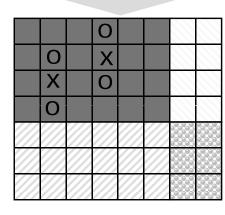
- Usually the block codes used in concatenated coding scheme are systematic
- \* Array within the heavy line box is stored in the interleaver array ( $k_2 \times n_1$  dimension)
- The composite code is much longer and more powerful
- → The data length is  $k_1 \times k_2$  and overall length is  $n_1 \times n_2$
- → This is called array or product code

- → Conventional decoding technique: decode inner code, then the outer
- It may not always be as effective as we might hope
- Assume both component codes are capable of correcting single errors only
- The 'O's are original received error patterns

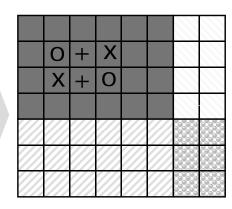


- •'+' indicates new errors added by outer decoder
- Some original errors ('o') are corrected

'x' indicates new errors added by inner decoder



Decode outer code

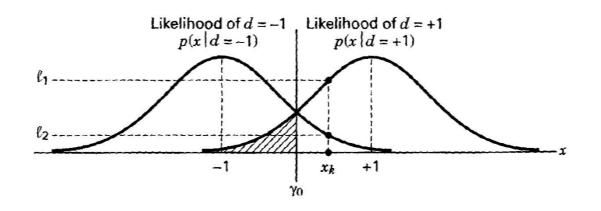


- → The final decoded codewords contained more errors than the original received codewords
- → If the output of the outer decoder are re-applied to the inner decoder it will detect that some errors remained
- → This is the principle of iterative decoder or 'turbo' principle

- However, for the above case the inner decoder may not be able to correct the errors
- The inner decoder needs additional information
- Hard decision made by demodulator and decoders destroy this information
- → Needs soft-in, soft-out (SISO) decoding

#### Turbo decoding

- Maximum a posteriori (MAP) decoding
- At the demodulator output:



- → Hypothesis  $H_1$ : likelihood is d = +1
- → Hypothesis  $H_2$ : likelihood is d = -1

- → At time k:
  - Demodulator output =  $x_k$
  - $-\ell_1 > \ell_2$ , or  $P(x_k | d=+1) > P(x_k | d=-1)$
- → Decision Rule:

$$P(d=+1 \mid X) > P(d=-1 \mid X)$$

$$H_2$$

→ Using Bayes' theorem:

$$P(d=+1 \mid x) p(x) > P(d=-1 \mid x) p(x)$$

$$H_{2}$$

$$H_{1}$$

$$p(x \mid d=+1) P(d=+1) > p(x \mid d=-1) P(d=-1)$$

$$H_{2}$$

→ Express in terms of a ratio:

$$\frac{P(d=+1|x)}{P(d=-1|x)} > 1 \quad or \quad \frac{p(x|d=+1)P(d=+1)}{p(x|d=-1)P(d=-1)} > 1 P(x|d=-1)P(d=-1) + 1 P(x|d=-1)P(d=-1)P(d=-1) + 1 P(x|d=-1)P(d=-1)P(d=-1)P(d=-1) + 1 P(x|d=-1)P(d=-$$

→ Take logarithm of LHS, we have Log-Likelihood Ratio (LLR):

$$L(d \mid x) = \log \left[ \frac{p(x \mid d = +1)P(d = +1)}{p(x \mid d = -1)P(d = -1)} \right]$$

Hence,

$$L(d \mid x) = \log \left[ \frac{p(x \mid d = +1)}{p(x \mid d = -1)} \right] + \log \left[ \frac{P(d = +1)}{P(d = -1)} \right]$$

$$L(x \mid d)$$

$$L(x \mid d)$$

- → L(x|a) is the LLR of the test statistic x obtained by measurement of x under the conditions d=+1 or d=-1.
- \* L(d) is the a priori LLR of the data bit d.

- → At the decoder output:
- → For systematic code, it has been shown that the LLR (soft decoder output) is

$$L_{d}(d|x) = L(d|x) + L_{e}(d|x)$$

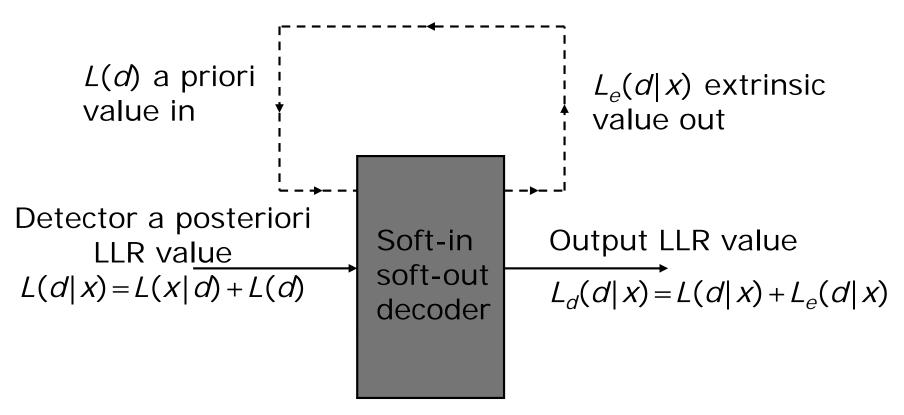
- → Where  $L_e(d|x) = \text{extrinsic LLR}$ , represents extra knowledge that is gleaned from the decoding process.
- → Hence,

$$L_{d}(d|x) = L(x|d) + L(d) + L_{e}(d|x)$$

- → The sign of L<sub>a</sub>(d|x) denotes hard decision:
  - -Positive: d = +1
  - -Negative: d = -1
- ▶ The magnitude of  $L_d(d|x)$  denotes the reliability of that decision.

- Iterative (turbo) decoding:
  - -Assume initially the binary data to be equally likely: set L(d) = 0
  - -Measure x and calculate  $\ell_1$  and  $\ell_2$ , then calculate L(x|d), hence L(d|x)
  - Decode and obtain  $L_e(d|x)$  from decoder output
  - Feedback  $L_e(d|x)$  and set  $L(d) = L_e(d|x)$
  - Re-calculate L(d|x) and decode

Feedback for the next iteration



Encoder output binary digits:

$d_1 = 1$	$d_2 = 0$	$p_{12} = 1$
$d_3 = 0$	$d_4 = 1$	$p_{34} = 1$
$p_{13} = 1$	$p_{24} = 1$	

Note:

$$d_{i} \oplus d_{j} = p_{ij}$$

$$d_{i} = d_{j} \oplus p_{ij}$$

$$d_{i} = d_{i} \oplus p_{ij}$$

→ Demodulator output (due to noise):

$x_1 = 0.75$	$x_2 = 0.05$	$x_{12} = 1.25$
$x_3 = 0.10$	$x_4 = 0.15$	$x_{34} = 1.00$
$x_{13} = 3.00$	$x_{24} = 0.50$	

Assume the following conversion:

$$0 \rightarrow -1$$
$$1 \rightarrow +1$$

→ Decoder input log-likelihood ratio, L(x|d)+0:

$$L(x|d) = \ln \left[ \frac{p(x|d=+1)}{p(x|d=-1)} \right]$$

$$= \ln \left[ \frac{1}{\sigma \sqrt{2\pi}} \exp \left[ -\frac{1}{2} \left( \frac{x-1}{\sigma} \right)^{2} \right] \right]$$

$$= \ln \left[ \frac{1}{\sigma \sqrt{2\pi}} \exp \left[ -\frac{1}{2} \left( \frac{x+1}{\sigma} \right)^{2} \right] \right]$$

$$= -\frac{1}{2} \left( \frac{x-1}{\sigma} \right)^2 + \frac{1}{2} \left( \frac{x+1}{\sigma} \right)^2 = \frac{2}{\sigma^2} x$$

→ To simplify, assume  $\sigma^2 = 1$ , then L(x|d) = L(x) = 2x

→ Hence,

$L(x_1) = 1.5$	$L(x_2) = 0.1$	$L(x_{12}) = 2.5$
$L(x_3) = 0.2$	$L(x_4) = 0.3$	$L(x_{34}) = 2.0$
$L(x_{13}) = 6.0$	$L(x_{24}) = 1.0$	

→ If hard decisions are made without decoding, d₂ and d₃ will be in errors.

→ LLR of modulo-2 sum of two bits:

$$L(b_{1} \oplus b_{2}) = \ln \left[ \frac{e^{L(b_{1})} + e^{L(b_{2})}}{1 + e^{L(b_{1})} e^{L(b_{2})}} \right]$$

$$\approx (-1) \times \text{sgn}[L(b_{1})] \times \text{sgn}[L(b_{2})]$$

$$\times \min(|L(b_{1})|, |L(b_{2})|)$$

\* Extrinsic LLR calculation, [to simplify notation,  $L_e(d_i|x_i) = L_e(d_i)$ ]:

$$L_{e}(\hat{d}_{i}) = L(\hat{d}_{j} \oplus \hat{p}_{ij})$$

$$\approx (-1) \cdot \operatorname{sgn}\left[L(\hat{d}_{j})\right] \cdot \operatorname{sgn}\left[L(\hat{p}_{ij})\right]$$

$$\cdot \min\left(L(\hat{d}_{j})\right|, \left|L(\hat{p}_{ij})\right|\right)$$

$$= (-1) \cdot \operatorname{sgn}\left[L(x_{j}) + L(d_{j})\right] \cdot \operatorname{sgn}\left[L(x_{ij})\right]$$

$$\cdot \min\left(L(x_{j}) + L(d_{j})\right|, \left|L(x_{ij})\right|\right)$$

- \*Assume  $L(\hat{p}_{ij}) = L(x_{ij})$  because  $p_{ij}$  depends on  $d_i$  and  $d_j$ .
- Decoding horizontally:

$$L_{eh}(\hat{a}_{1}) = (-1) \cdot \text{sgn}[L(x_{2}) + L(a_{2})] \cdot \text{sgn}[L(x_{12})]$$

$$\cdot \min(|L(x_{2}) + L(d_{2})|, |L(x_{12})|)$$

$$= (-1) \cdot \text{sgn}[0.1 + 0] \cdot \text{sgn}[2.5] \cdot (0.1)$$

$$= (-1) \cdot (+1) \cdot (+1) \cdot (0.1)$$

$$= -0.1 = new L(a_{1})$$

#### → Similarly:

$$-L_{eh}(\hat{d}_{2}) = -\operatorname{sgn}(1.5+0) \cdot \operatorname{sgn}(2.5) \cdot (1.5)$$

$$= -1.5 = \operatorname{new} L(d_{2})$$

$$-L_{eh}(\hat{d}_{3}) = -\operatorname{sgn}(0.3+0) \cdot \operatorname{sgn}(2.0) \cdot (0.3)$$

$$= -0.3 = \operatorname{new} L(d_{3})$$

$$-L_{eh}(\hat{d}_{4}) = -\operatorname{sgn}(0.2+0) \cdot \operatorname{sgn}(2.0) \cdot (0.2)$$

$$= -0.2 = \operatorname{new} L(d_{4})$$

Decoding vertically:

$$-L_{ev}(\overset{\wedge}{d}_{1}) = 0.1 = \text{new } L(d_{1})$$
  
 $-L_{ev}(\overset{\wedge}{d}_{2}) = -0.1 = \text{new } L(d_{2})$   
 $-L_{ev}(\overset{\wedge}{d}_{3}) = -1.4 = \text{new } L(d_{3})$   
 $-L_{ev}(\overset{\wedge}{d}_{4}) = 1.0 = \text{new } L(d_{4})$ 

→ Final decoder output log-likelihood ratio after first iteration:

$$L_d(d_i|x_i) = L(x_i) + L_{eh}(d_i) + L_{ev}(d_i)$$

Hence,

-Final 
$$L_d(d_1|x_1) = 1.5 - 0.1 + 0.1 = 1.5$$

-Final 
$$L_o(d_2|x_2) = 0.1 - 1.5 - 0.1 = -1.5$$

-Final 
$$L_d(d_3|x_3) = 0.2 - 0.3 - 1.4 = -1.5$$

-Final 
$$L_d(d_4|x_4) = 0.3 - 0.2 + 1.0 = 1.1$$

→ Final decoder output after 1<sup>st</sup> iteration:

$L_{d}(d_{1} x_{1})=1.5$	$L_d(d_2 x_2) = -1.5$
$L_d(d_3 x_3) = -1.5$	$L_d(d_4 X_4) = 1.1$

- → After 1<sup>st</sup> iteration, it is sufficient to yield correct hard decision outputs for  $d_3$  and  $d_4$ .
- → Let's see if 2<sup>nd</sup> iteration can improve the reliability or higher confidence.
- → For  $2^{nd}$  iteration, repeat horizontal and vertical decodings with new  $L(d_i)$  values.

Horizontal decoding:

$$-L_{eh}(\hat{d}_{1}) = -\operatorname{sgn}(0.1-0.1) \cdot \operatorname{sgn}(2.5) \cdot (0)$$

$$= 0 = \operatorname{new} L(d_{1})$$

$$-L_{eh}(\hat{d}_{2}) = -\operatorname{sgn}(1.5+0.1) \cdot \operatorname{sgn}(2.5) \cdot (1.6)$$

$$= -1.6 = \operatorname{new} L(d_{2})$$

$$-L_{eh}(\hat{d}_{3}) = -\operatorname{sgn}(0.3+1.0) \cdot \operatorname{sgn}(2.0) \cdot (1.3)$$

$$= -1.3 = \operatorname{new} L(d_{3})$$

$$-L_{eh}(\hat{d}_{4}) = -\operatorname{sgn}(0.2-1.4) \cdot \operatorname{sgn}(2.0) \cdot (|-1.2|)$$

$$= 1.2 = \operatorname{new} L(d_{4})$$

#### Vertical decoding:

$$-L_{ev}(d_1) = 1.1 = \text{new L}(d_1)$$
  
 $-L_{ev}(d_2) = -1.0 = \text{new L}(d_2)$   
 $-L_{ev}(d_3) = -1.5 = \text{new L}(d_3)$   
 $-L_{ev}(d_4) = 1.0 = \text{new L}(d_4)$ 

#### → Final LLR after 2<sup>nd</sup> iteration:

$$-L_d(d_1|x_1) = 1.5 + 0 + 1.1 = 2.6$$

$$-L_d(d_2|x_2) = 0.1 - 1.6 - 1.0 = -2.5$$

$$-L_d(d_3|x_3) = 0.2 - 1.3 - 1.5 = -2.6$$

$$-L_d(d_4|x_4) = 0.3 + 1.2 + 1.0 = 2.5$$

→ Hence final decoder output after 2<sup>nd</sup> iteration:

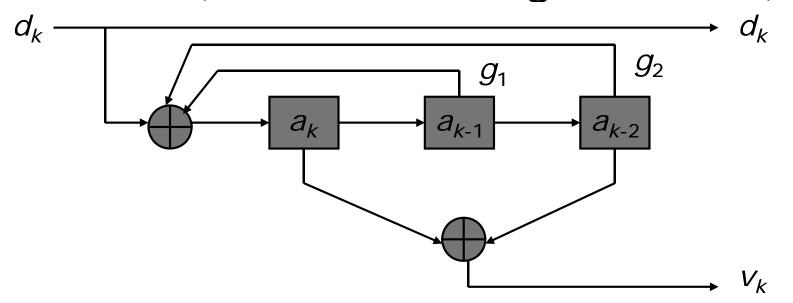
$L_d(d_1 x_1)=2.6$	$L_d(d_2 x_2) = -2.5$
$L_d(d_3 x_3) = -2.6$	$L_d(d_4 x_4) = 2.5$

• We see that the level of decision confidence increased.

#### **Turbo-Convolutional Codes**

- Invented by C. Berrou, A. Glavieux and P. Thitimajshima in 1993.
- → It is the original Turbo Codes.
- It is a parallel (not serial) concatenated recursive systematic convolutional (RSC) codes.
- → Recursive codes are used because they give better performance than the best non-systematic codes at all E<sub>b</sub>/N<sub>0</sub> for high code rates.

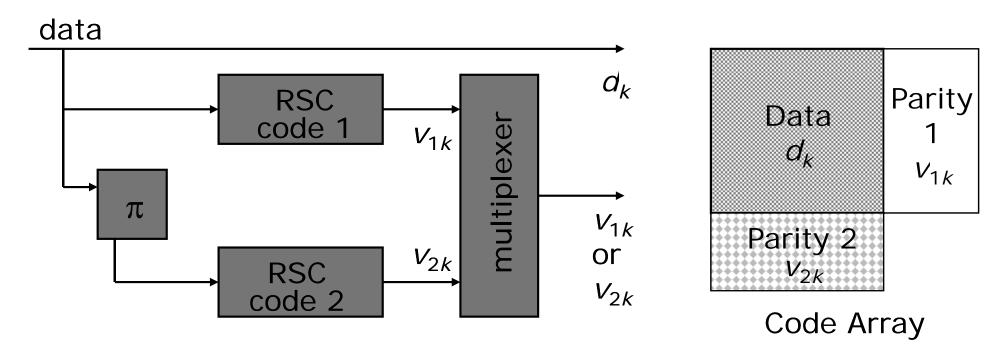
\* RSC code (constraint length, K = 3):



Mathematical representation:

$$a_k = d_k + \sum_{i=1}^{K-1} g_i a_{k-i} \mod 2$$

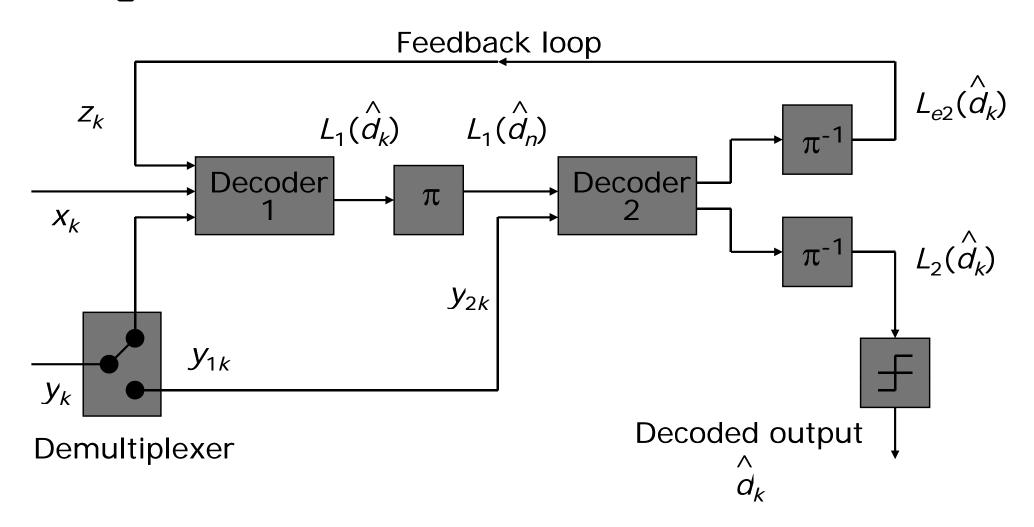
Structure of parallel concatenation:



Code array is the same as serial concatenation, except no 'check-oncheck'.

- Iterative decoding:
  - Viterbi Algorithm is an optimal decoding method for minimizing the probability of sequence error, not bit error.
  - Output of Viterbi is hard decision of a sequence of bits.
  - For iterative decoding, we need softdecision output for each decoded bit.
- → Need to use Bahl Algorithm (beyond the scope of this course).

Original Berrou's Feedback Decoder:



BER performance

