

- Soft decision decoding (or equivalently binary-input additive white Gaussian channel)
 - The error of Ungerboeck codes (at high SNR) is dominated by the “two code words” whose pairwise Euclidean distance is equal to d_{free} . (This d_{free} represents Euclidean distance, not the Hamming distance defined previously.)

⇒ Equivalently $x_j = s_{j,m} + w_j$ for $j = 1, \dots, N$

$$\text{where } \|\mathbf{s}_0 - \mathbf{s}_1\|^2 = \sum_{j=1}^N (s_{j,0} - s_{j,1})^2 = d_{\text{free}}^2$$

⇒ $\hat{m} = \arg \max \{P(\mathbf{x} | \mathbf{s}_0), P(\mathbf{x} | \mathbf{s}_1)\}$

$$\Rightarrow \hat{m} = \arg \max \left\{ \prod_{j=1}^N e^{-(x_j - s_{j,0})^2 / 2\sigma^2}, \prod_{j=1}^N e^{-(x_j - s_{j,1})^2 / 2\sigma^2} \right\}$$

$$\Rightarrow \|\mathbf{x} - \mathbf{s}_0\|^2 \underset{\mathbf{s}_1}{\overset{\mathbf{s}_0}{\leq}} \|\mathbf{x} - \mathbf{s}_1\|^2 \quad \mathbf{x} \begin{cases} \mathcal{N}(\mathbf{s}_0, \sigma^2 \mathbb{I}) & \mathbf{s}_0 \text{ transmitted} \\ \mathcal{N}(\mathbf{s}_1, \sigma^2 \mathbb{I}) & \mathbf{s}_1 \text{ transmitted} \end{cases}$$

- (Approximate) Error probability

- Based on the decision rule $\|\mathbf{x} - \mathbf{s}_0\|^2 \underset{\mathbf{s}_1}{\overset{\mathbf{s}_0}{\leq}} \|\mathbf{x} - \mathbf{s}_1\|^2$

Dominant pairwise error

$$\begin{aligned} &= P(\mathbf{s}_0 \text{ transmitted}) P(\|\mathbf{x} - \mathbf{s}_0\|^2 > \|\mathbf{x} - \mathbf{s}_1\|^2 | \mathbf{s}_0 \text{ transmitted}) \\ &\quad + P(\mathbf{s}_1 \text{ transmitted}) P(\|\mathbf{x} - \mathbf{s}_0\|^2 < \|\mathbf{x} - \mathbf{s}_1\|^2 | \mathbf{s}_1 \text{ transmitted}) \\ &= P(\|\mathbf{x} - \mathbf{s}_0\|^2 > \|\mathbf{x} - \mathbf{s}_1\|^2 | \mathbf{s}_0 \text{ transmitted}) \\ &= P(\|\mathbf{w}\|^2 > \|\mathbf{w} + \mathbf{s}_0 - \mathbf{s}_1\|^2), \text{ where } \mathbf{x} = \mathbf{s}_0 + \mathbf{w} \\ &= P\left(\langle \mathbf{w}, \mathbf{s}_0 - \mathbf{s}_1 \rangle < -\frac{1}{2} \|\mathbf{s}_0 - \mathbf{s}_1\|^2\right) \quad \begin{matrix} \langle \mathbf{w}, \mathbf{s}_0 - \mathbf{s}_1 \rangle \\ \sim \mathcal{N}(0, \|\mathbf{s}_0 - \mathbf{s}_1\|^2 \sigma^2) \end{matrix} \\ &= \Phi\left(\frac{-\frac{1}{2} \|\mathbf{s}_0 - \mathbf{s}_1\|^2 - 0}{\|\mathbf{s}_0 - \mathbf{s}_1\| \sigma}\right) = \Phi\left(-\frac{d_{\text{free}}}{2\sigma}\right) \leq \exp\{-d_{\text{free}}^2 / (4N_0)\} \end{aligned}$$

$$\Phi(-x) = \frac{1}{2} \operatorname{erfc}\left(\frac{x}{\sqrt{2}}\right) \leq \frac{1}{x\sqrt{2\pi}} e^{-x^2/2} \leq e^{-x^2/2} \text{ for } x > 1/\sqrt{2\pi} \quad \sigma^2 = N_0/2$$

□ Asymptotic Coding gain (asymptotic = at high SNR) G_a

■ The performance gain due to coding (i.e., the performance gain of coded system against uncoded system)

$$\text{Uncoded } \exp \left\{ -d_{\text{ref}}^2 / (4N_0) \right\}$$

$$\text{Coded system } \exp \left\{ -d_{\text{free}}^2 / (4N_0) \right\}$$

$$G_a = 10 \log_{10} \left(\frac{d_{\text{free}}^2}{d_{\text{ref}}^2} \right)$$

□ 4-state Ungerboeck code

■ Its code rate is 2 bits/symbol; hence, it should be compared with uncoded QPSK.

$$d_0 = 2 \sin \left(\frac{\pi}{8} \right) = \sqrt{2} - \sqrt{2}$$

$$d_1 = \sqrt{2}$$

$$d_2 = 2$$

Signal number

00

10

01

11

0

2

1

3

$$\frac{d_{\text{free}}}{d_{\text{ref}}} = \frac{d_2}{d_1} = \frac{2}{\sqrt{2}} \Rightarrow G_a = 10 \log_{10} \left(\frac{d_{\text{free}}^2}{d_{\text{ref}}^2} \right) = 3 \text{ dB}$$

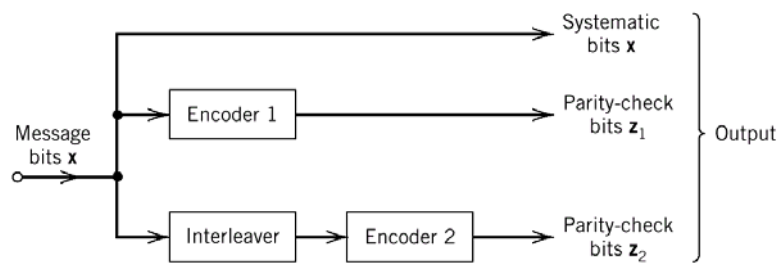
Number of states	4	8	16	32	64	128	256	512
Coding gain (dB)	3	3.6	4.1	4.6	4.8	5	5.4	5.7

□ Final note

- Asymptotic coding gain of Ungerboeck codes increases as the number of states grows.
- To have a 6 dB coding gain, we may need an Ungerboeck code with a large number of states.

10.8 Turbo codes

□ Structure of turbo code encoder

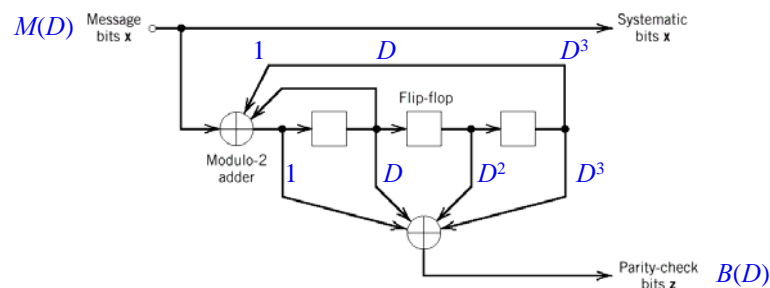


10.8 Turbo codes

- Basic consideration for such design
 - Add an interleaver to tie together distant bits.
 - Use *recursive* systematic convolutional (RSC) codes to make the **internal state depend on the past outputs**.
 - Use recursive *systematic* convolutional (RSC) codes to make the **turbo-like iterative decoding possible**.
 - RSC code may suffer *catastrophic error propagation* (one single output error produces an infinite number of parity errors).
 - Use *short constraint-length* RSC codes to **reduce to decoding burden in each decoding iteration**.

10.8 Turbo codes

- Example 10.8 Eight-state RSC (constituent) encoder



$$g(D) = \left[1, \frac{1 + D + D^2 + D^3}{1 + D + D^3} \right]$$

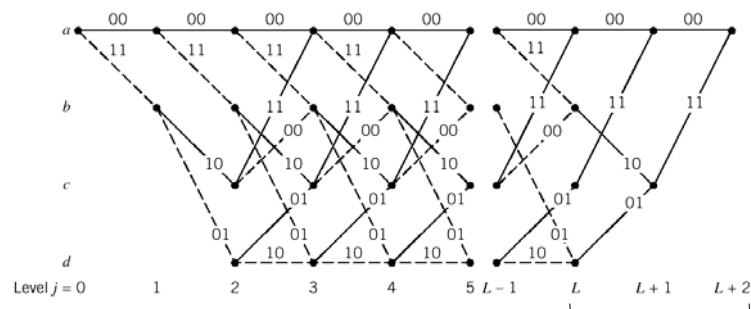
10.8 Turbo codes

$$\frac{B(D)}{M(D)} = \frac{1 + D + D^2 + D^3}{1 + D + D^3}$$

$$\Rightarrow b_i = m_i + m_{i-1} + m_{i-2} + m_{i-3} - b_{i-1} - b_{i-3}$$

□ Remark 1: Zero tail bits

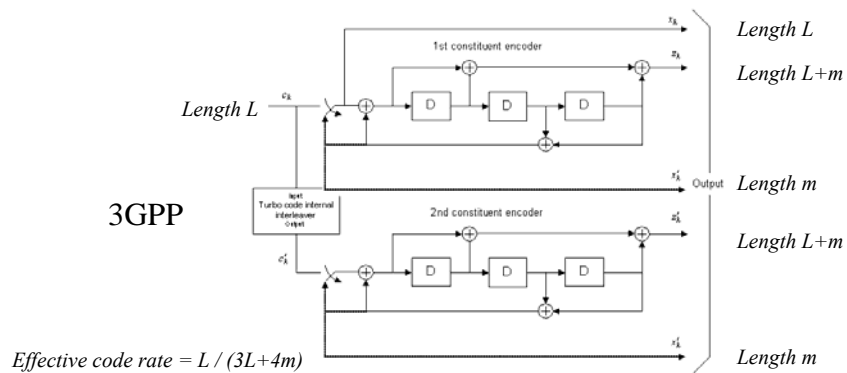
- With the pseudo-random interleaver, the zero tail bits for the first encoder may not appear to be the tail bits of the second encoder.



Appended two zeros to clear the shift-register contents

□ Remark 1: Zero tail bits (continue)

- With a more careful design, dual clearing of the two encoder register contents can be achieved, which results in considerable performance improvement at medium to high signal-to-noise ratios.



□ Remark 2: Punctured convolution code.

- Let L be the number of information bits.
- Each $(2, 1)$ constituent encoder will generate approximately L parity-check bits.
- With two constituent encoders, the system transmits L information bits and $2L$ parity-check bits, which reduces the code rate to $1/3$.

□ Remark 2: Punctured convolution code (continue)

- To improve the code rate, one can “puncture” half of the parity-check bits generated by each constituent encoder.
- With two constituent encoders, the system transmits L information bits and $L = (L/2) * 2$ parity-check bits, which reduces the code rate to $1/2$.
- At the decoder side, since we exactly know that “no transmission” is performed in those punctured positions, we can directly “nullify” (i.e., make them zero) the corresponding received scalars.

■ For example,

x_1 and x_2 are information bits.

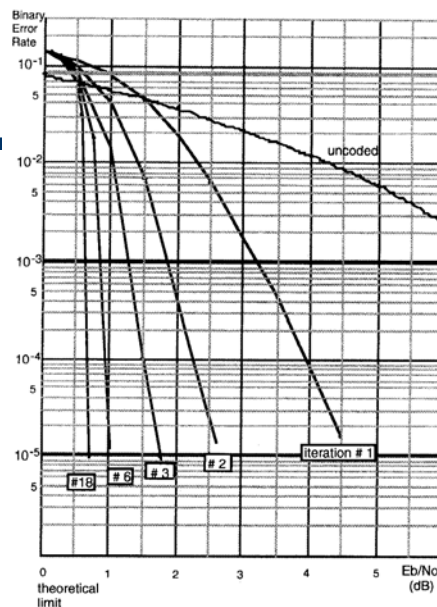
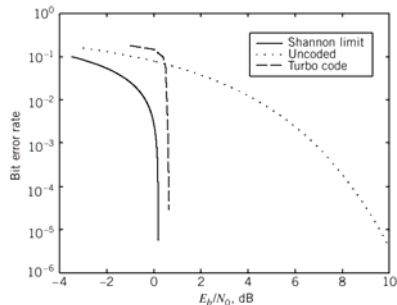
$$[r_1 \ r_2 \ r_3 \ r_5] = [x_1 \ x_2 \ x_3 \ x_5] + [w_1 \ w_2 \ w_3 \ w_5]$$

Parity-check bits x_4 and x_6 are punctured.

Decoder decodes x_1 and x_2 based on $[r_1 \ r_2 \ r_3 \ 0 \ r_5 \ 0]$.

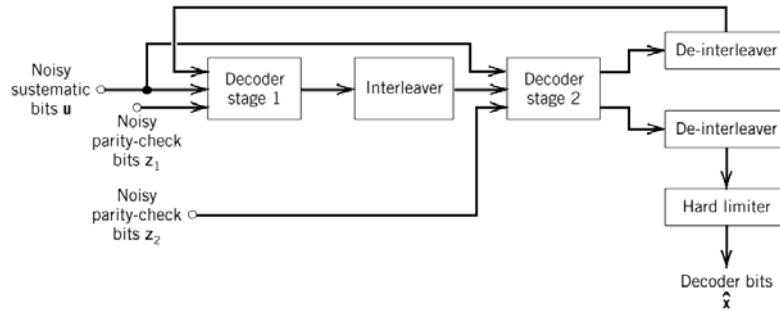
10.8 Turbo codes

□ Performance of Turbo codes



10.8 Turbo codes

□ Turbo decoder



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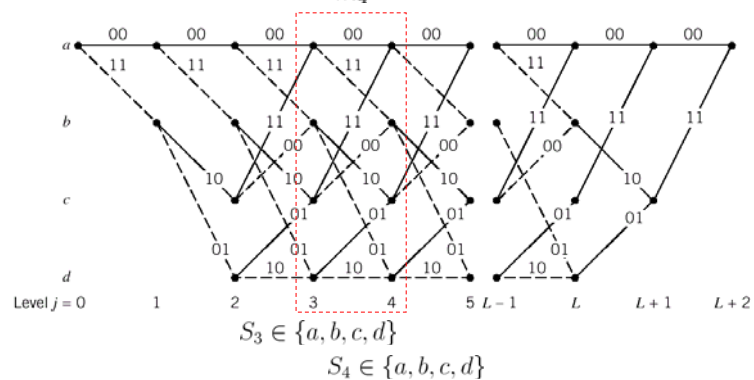
Chapter 10-115

□ Turbo component decoder (BCJR algorithm or log-MAP algorithm)

- Only apply to decode a code whose present state and present output are a function of the past state and current input bit.

Set of transition corresponding to symbol 0 : $\mathcal{B}_{3,4}(0) = \{(a, a), (b, c), (c, a), (d, c)\}$

Set of transition corresponding to symbol 1 : $\mathcal{B}_{3,4}(1) = \{(a, b), (b, d), (c, b), (d, d)\}$



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Chapter 10-116

- It minimizes the bit error directly rather than word error.

$$P(m_4 = 0 | \mathbf{r}) = P((S_3, S_4) \in \mathcal{B}_{3,4}(0) | \mathbf{r})$$

Left-hand side = The probability of the 4th message bit given that the receiver receives \mathbf{r} .

Right-hand side = The probability of the encoder going through state S_3 and state S_4 in $\mathcal{B}_{3,4}(0)$ given that the receiver receives \mathbf{r} .

$$\begin{aligned} \Rightarrow l(4) &= \log \frac{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(1)} P(S_3, S_4 | \mathbf{r})}{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(0)} P(S_3, S_4 | \mathbf{r})} \\ &= \log \frac{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(1)} P(S_3, S_4, \mathbf{r})}{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(0)} P(S_3, S_4, \mathbf{r})} \end{aligned}$$

My derivation is different from the text, which is based on the original work.

$$\begin{aligned} P(S_3, S_4, \mathbf{r}) &= P(S_3, S_4, r_1^6, r_7^8, r_9^N) \\ &= P(r_9^N | S_3, S_4, r_1^6, r_7^8) P(S_3, S_4, r_1^6, r_7^8) \\ &= \underbrace{P(r_9^N | S_4)}_{\beta(S_4)} P(S_3, S_4, r_1^6, r_7^8) \end{aligned}$$

$(S_3, r_1^6, r_7^8) \rightarrow S_4 \rightarrow r_9^N$
forms a Markov chain

$$\begin{aligned} P(S_3, S_4, r_1^6, r_7^8) &= P(S_3, r_1^6) P(S_4, r_7^8 | S_3, r_1^6) \\ &= \underbrace{P(S_3, r_1^6)}_{\alpha(S_3)} \underbrace{P(S_4, r_7^8 | S_3)}_{\gamma(S_3, S_4)} \end{aligned}$$

$$\Rightarrow l(4) = \log \frac{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(1)} \alpha(S_3) \beta(S_4) \gamma(S_3, S_4)}{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(0)} \alpha(S_3) \beta(S_4) \gamma(S_3, S_4)}$$

$$\begin{aligned}
\alpha(S_3) &= P(S_3, r_1^6) \\
&= \sum_{S_2 \in \{a,b,c,d\}} P(S_2, S_3, r_1^4, r_5^6) \\
&= \sum_{S_2 \in \{a,b,c,d\}} P(S_2, r_1^4) P(S_3, r_5^6 | S_2, r_1^4) \\
&= \sum_{S_2 \in \{a,b,c,d\}} P(S_2, r_1^4) P(S_3, r_5^6 | S_2) \\
&= \sum_{S_2 \in \{a,b,c,d\}} \alpha(S_2) \gamma(S_2, S_3)
\end{aligned}$$

$$\text{Initial value } \alpha(S_0) = P(S_0, r_1^0) = P(S_0) = \begin{cases} 1, & S_0 = a; \\ 0, & S_0 = b; \\ 0, & S_0 = c; \\ 0, & S_0 = d \end{cases}$$

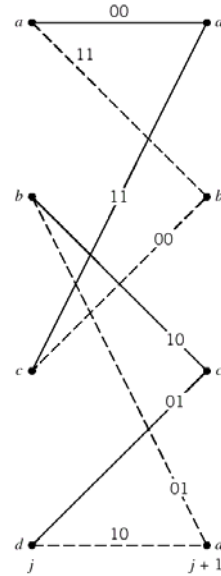
$$\begin{aligned}
\beta(S_4) &= P(r_9^N | S_4) \\
&= \sum_{S_5 \in \{a,b,c,d\}} P(S_5, r_9^{10}, r_{10}^N | S_4) \\
&= \sum_{S_5 \in \{a,b,c,d\}} P(r_{10}^N | S_4, S_5, r_9^{10}) P(S_5, r_9^{10} | S_4) \\
&= \sum_{S_5 \in \{a,b,c,d\}} P(r_{10}^N | S_5) P(S_5, r_9^{10} | S_4) \\
&= \sum_{S_5 \in \{a,b,c,d\}} \beta(S_5) \gamma(S_4, S_5)
\end{aligned}$$

$$\text{Initial value } \beta(S_{L+m}) = \begin{cases} 1, & S_{L+m} = a; \\ 0, & S_{L+m} = b; \\ 0, & S_{L+m} = c; \\ 0, & S_{L+m} = d \end{cases}$$

Note that the turbo codes use *systematic* code; hence, one of the code bit should be the same as the message bit. The example in the next slide is simply taken from the textbook for convenience (so that I don't have to draw a figure myself.)

$$\begin{aligned}
\gamma(S_3, S_4) &= P(S_4, r_7^8 | S_3) \\
&= P(r_7^8 | S_3, S_4) P(S_4 | S_3) \\
&= \underbrace{P(r_7^8 | x_7^8(S_3, S_4))}_{\text{channel}} \underbrace{P(S_4 | S_3)}_{\text{prior}} = \underbrace{P(r_7^8 | m_4, x_8(S_3, S_4))}_{\text{channel}} \underbrace{P(S_4 | S_3)}_{\text{prior}} \\
&= \underbrace{P(r_7 | m_4)}_{\text{systematic}} \underbrace{P(r_8 | x_8(S_3, S_4))}_{\text{parity}} \underbrace{P(S_4 | S_3)}_{\text{prior}}
\end{aligned}$$

$$P(S_4|S_3) = \begin{cases} P(a|a) = P(m_4 = 0) \\ P(b|a) = P(m_4 = 1) \\ P(c|a) = 0 \\ P(d|a) = 0 \\ P(a|b) = 0 \\ P(b|b) = 0 \\ P(c|b) = P(m_4 = 0) \\ P(d|b) = P(m_4 = 1) \\ P(a|c) = P(m_4 = 0) \\ P(b|c) = P(m_4 = 1) \\ P(c|c) = 0 \\ P(d|c) = 0 \\ P(a|d) = 0 \\ P(b|d) = 0 \\ P(c|d) = P(m_4 = 0) \\ P(d|d) = P(m_4 = 1) \end{cases}$$



Let $\tilde{\gamma}(S_3, S_4) = \underbrace{P(r_8|x_8(S_3, S_4))}_{\text{parity}}$

$$\begin{aligned} \Rightarrow l(4) &= \log \frac{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(1)} \alpha(S_3)\beta(S_4)\tilde{\gamma}(S_3, S_4)P(S_4|S_3)}{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(0)} \alpha(S_3)\beta(S_4)\tilde{\gamma}(S_3, S_4)P(S_4|S_3)} \\ &= \log \frac{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(1)} \alpha(S_3)\beta(S_4)\tilde{\gamma}(S_3, S_4)P(m_4 = 1)P(r_7|1)}{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(0)} \alpha(S_3)\beta(S_4)\tilde{\gamma}(S_3, S_4)P(m_4 = 0)P(r_7|0)} \\ &= \underbrace{\log \frac{P(m_4 = 1)}{P(m_4 = 0)}}_{\substack{\text{a priori } l_{in} \\ \text{(intrinsic)}}} + \underbrace{\log \frac{P(r_7|1)}{P(r_7|0)}}_{\text{systematic}} + \underbrace{\log \frac{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(1)} \alpha(S_3)\beta(S_4)\tilde{\gamma}(S_3, S_4)}{\sum_{(S_3, S_4) \in \mathcal{B}_{3,4}(0)} \alpha(S_3)\beta(S_4)\tilde{\gamma}(S_3, S_4)}}_{\text{extrinsic } l_{ex}} \end{aligned}$$

Step 1: With $l_{\text{in}}(m_j)$ known, we can compute

$$P(m_j = 0) = \frac{1}{1 + \exp\{l_{\text{in}}(m_j)\}} \text{ and } P(m_j = 1) = \frac{\exp\{l_{\text{in}}(m_j)\}}{1 + \exp\{l_{\text{in}}(m_j)\}}$$

for each $1 \leq j \leq L$. We can in turn compute $\gamma(S_j, S_{j+1})$ for all $(S_j, S_{j+1}) \in \{a, b, c, d\}^2$ and for each $0 \leq j \leq j + m$.

Step 2: With γ available, we can recursively compute α and β in the forward and backward fashion, respectively.

Step 3: With α , β and γ ready, we can compute

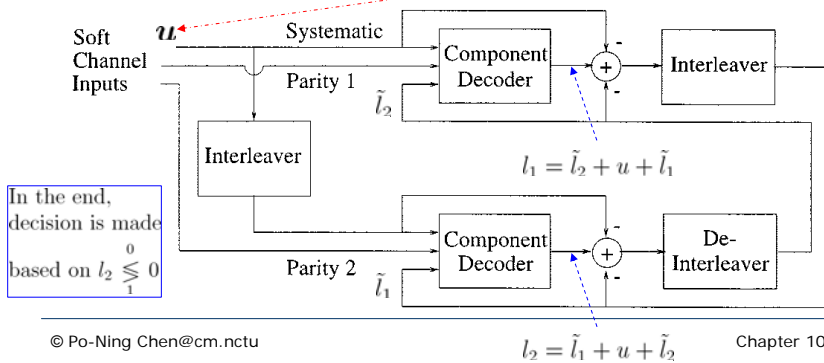
$$l(j) = \log \frac{P(m_j = 1|\mathbf{r})}{P(m_j = 0|\mathbf{r})} = \log \frac{\sum_{(S_{j-1}, S_j) \in \mathcal{B}_{j-1, j}(1)} \alpha(S_{j-1})\beta(S_j)\gamma(S_{j-1}, S_j)}{\sum_{(S_{j-1}, S_j) \in \mathcal{B}_{j-1, j}(0)} \alpha(S_{j-1})\beta(S_j)\gamma(S_{j-1}, S_j)}$$

Step 4: Calculate $l_{\text{ex}}(j) = l(j) - \log \frac{P(r_{2j-1}|1)}{P(r_{2j-1}|0)} - l_{\text{in}}(j)$.

Observe $l(j) = l_{\text{in}}(j) + \log \frac{P(r_{2j-1}|1)}{P(r_{2j-1}|0)} + l_{\text{ex}}(j)$.

Intuition: Recursion between the two.
(The a priori information should come from an "independent source" and not to reuse the same information more than once.)

$$\begin{aligned} u_j &= \log \frac{P(r_{2j-1}|1)}{P(r_{2j-1}|0)} \\ &= \log \frac{\exp\left\{-\frac{(r_{2j-1}-(+1))^2}{2\sigma^2}\right\}}{\exp\left\{-\frac{(r_{2j-1}-(-1))^2}{2\sigma^2}\right\}} \\ &= \frac{2}{\sigma^2} r_{2j-1} \end{aligned}$$



- Turbo coding, although quite impressive in performance, is designed based on “good” intuition.
- For example, Berrou, Glaviexu and Thitimajshima’s wrote in their 1993 ICC paper (the first paper for turbo coding technique) that

□ ... for very low SNRs, the BER can sometimes increase during the iterative decoding process. In order to overcome this effect, the extrinsic information \tilde{l}_1 (resp. \tilde{l}_2) has been divided by $(1+\theta|l_1|)$ (resp. $(1+\theta|l_2|)$). θ acts as a stability factor and its value of 0.15 was adopted after several simulation tests at $E_b/N_0 = 0.7$ dB.... [1, pp. 1270]

[1] Claude Berrou and Alain Glaviexu, “Near optimal error correcting coding and decoding: Turbo-codes,” IEEE Transactions on Communications, vol. 44, no. 10, pp. 1261-1271, October 1996.

10.9 Computer experiment: Turbo coding

- Read by yourself.

10.10 Low-density parity-check codes

- LDPC codes (also known as Gallager codes) are also iteratively decodable.

- Its advantage over turbo coding technique is
 - Absence of low-weight code words.
 - With careless interleaver design, the turbo code may have low weight codewords, which is the main cause for error floor.
 - Iterative decoding of lower complexity.

10.10 Low-density parity-check codes

- In notations, a LDPC code is usually denoted by three tuple (n, t_c, t_r) .
 - n = block length
 - t_c = number of 1s in each column (of $(n-k)$ bits)
 - t_r = number of 1s in each row (of n bits)

 - It is not necessary to specify k since

$$(\# \text{ of 1s}) = nt_c = (n - k)t_r \Rightarrow \frac{t_c}{t_r} = 1 - \frac{k}{n}$$

□ In terminologies,

$$\mathbf{H}_{(n-k) \times n} \mathbf{G}_{n \times k}^T = \mathbf{0}_{(n-k) \times k}$$

$$\mathbf{c}_{1 \times n} \mathbf{H}_{n \times (n-k)}^T = \mathbf{m}_{1 \times k} \mathbf{G}_{k \times n} \mathbf{H}_{n \times (n-k)}^T = \mathbf{0}_{1 \times (n-k)}$$

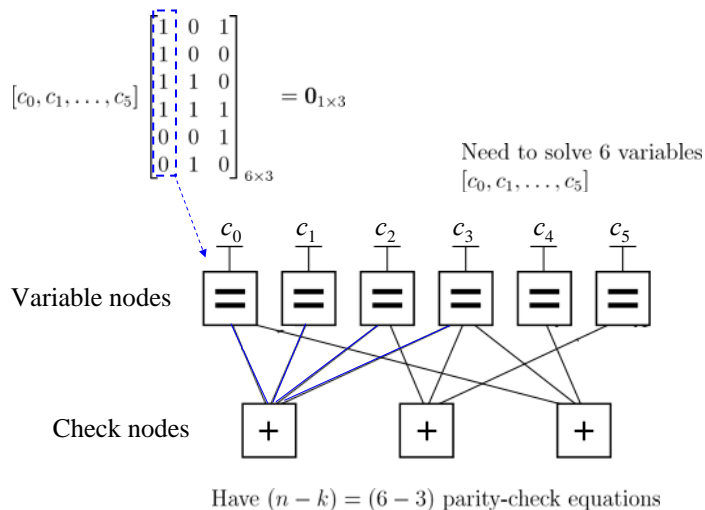
$$\mathbf{c}_{1 \times n} = \mathbf{m}_{1 \times k} \mathbf{G}_{k \times n}$$

■ Example. $(n, t_c, t_r) = (10, 3, 5)$.

$$\frac{k}{n} = 1 - \frac{t_c}{t_r} = 1 - \frac{3}{5} = \frac{2}{5}$$

$$[c_0, c_1, \dots, c_9] \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 0 \\ 1 & \ddots & & & & \\ & & \ddots & & & \\ 1 & \ddots & & & & \\ \vdots & & & & & \end{bmatrix}_{10 \times (10-4)} = \mathbf{0}_{1 \times (10-4)}$$

■ Forney's factor graph (Bipartite graph)



10.10 Low-density parity-check codes

- How to find the generator matrix for a given parity-check matrix?

- Presumption: Systematic LDPC code

$$\mathbf{c} = [\mathbf{b} : \mathbf{m}]$$

$$\text{where } \begin{cases} \mathbf{b} \text{ are parity-check bits} \\ \mathbf{m} \text{ are message bits} \end{cases}$$

$$\mathbf{H}_{(n-k) \times n}^T = \begin{bmatrix} \mathbf{H}_1 \\ \cdots \\ \mathbf{H}_2 \end{bmatrix} \Rightarrow [\mathbf{b} : \mathbf{m}] \begin{bmatrix} \mathbf{H}_1 \\ \cdots \\ \mathbf{H}_2 \end{bmatrix} \Rightarrow \mathbf{b}\mathbf{H}_1 + \mathbf{m}\mathbf{H}_2 = \mathbf{0}$$

The generator matrix of a systematic code must of the shape

$$\mathbf{G} = \begin{bmatrix} \mathbf{P}_{k \times (n-k)} & \vdots & \mathbf{I}_k \end{bmatrix} \Rightarrow \mathbf{m}\mathbf{P} = \mathbf{b}$$

This concludes to:

$$\mathbf{m}\mathbf{P}\mathbf{H}_1 + \mathbf{m}\mathbf{H}_2 = \mathbf{0} \Rightarrow \mathbf{P} = \mathbf{H}_2\mathbf{H}_1^{-1} \Rightarrow \mathbf{G} = \begin{bmatrix} \mathbf{H}_2^{-1}\mathbf{H}_1 & \vdots & \mathbf{I}_k \end{bmatrix}$$

10.10 Low-density parity-check codes

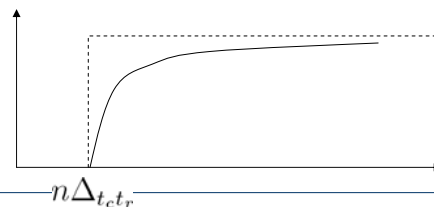
□ Remarks

- Low-density parity-check code gets its name since the number of 1s in each row and column is small (low-density).
- If the number of 1s in each row and also in each column is fixed, the LDPC code is said to be *regular*.
- Under regularity, the inverse matrix of \mathbf{H}_1 may be difficult to make to exist.
- Hence, some “manipulation” or even allowing some “irregularity” is sometimes necessary.

10.10 Low-density parity-check codes

□ Minimum distance of LDPC codes

- By uniformly selecting code word pairs, the pairwise distance becomes a random variable, for which the cumulative distribution function (cdf) can be empirically plotted.
- It is shown that this cdf can be overbounded by a unit step function as shown below.

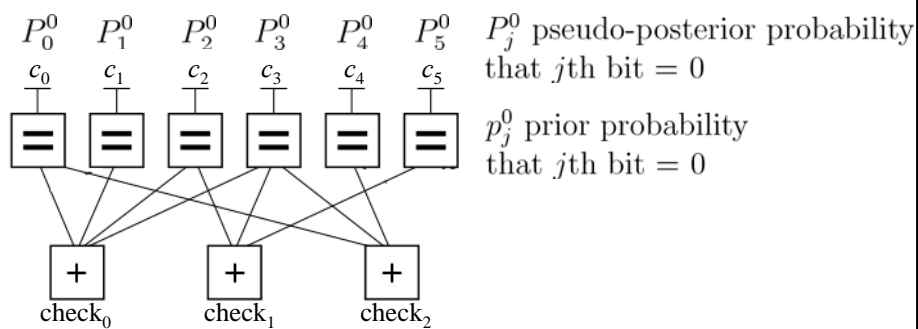


10.10 Low-density parity-check codes

t_c	t_r	Code rate k/n	$\Delta_{t_c t_r}$
5	6	0.167	0.255
4	5	0.2	0.210
3	4	0.25	0.122
4	6	0.333	0.129
3	5	0.4	0.044
3	6	0.5	0.023

10.10 Low-density parity-check codes

□ Probabilistic decoding of LDPC codes



Based on

$$p_j^0 = P(c_j = 0|r_j)$$

$$= \frac{P(c_j = 0)P(r_j|c_j = 0)}{P(c_j = 0)P(r_j|c_j = 0) + P(c_j = 1)P(r_j|c_j = 1)}$$

to find the most probable \mathbf{c} such that $\mathbf{c}\mathbf{H}^T = \mathbf{0}$
through recursion.

$$\begin{bmatrix} Q_{0,0}^0 & Q_{1,0}^0 & Q_{2,0}^0 \\ Q_{0,1}^0 & Q_{1,1}^0 & Q_{2,1}^0 \\ Q_{0,2}^0 & Q_{1,2}^0 & Q_{2,2}^0 \\ Q_{0,3}^0 & Q_{1,3}^0 & Q_{2,3}^0 \\ Q_{0,4}^0 & Q_{1,4}^0 & Q_{2,4}^0 \\ Q_{0,5}^0 & Q_{1,5}^0 & Q_{1,5}^0 \end{bmatrix} \Leftrightarrow \begin{bmatrix} P_{0,0}^0 & P_{1,0}^0 & P_{2,0}^0 \\ P_{0,1}^0 & P_{1,1}^0 & P_{2,1}^0 \\ P_{0,2}^0 & P_{1,2}^0 & P_{2,2}^0 \\ P_{0,3}^0 & P_{1,3}^0 & P_{2,3}^0 \\ P_{0,4}^0 & P_{1,4}^0 & P_{2,4}^0 \\ P_{0,5}^0 & P_{1,5}^0 & P_{1,5}^0 \end{bmatrix} \quad \begin{matrix} \boxed{\begin{matrix} 1 & 0 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{matrix}}_{6 \times 3} = \mathbf{0}_{1 \times 3} \end{matrix}$$

Initialization:

$$P_{i,j}^0 = p_j^0$$

Horizontal step:

$$Q_{i,j}^0 = \frac{1}{2} \left(1 + \prod_{k \in \text{Check}(i) - \{j\}} (2P_{i,k}^0 - 1) \right)$$

Vertical step:

$$P_{i,j}^0 = \frac{p_j^0 \prod_{k \in \text{Bit}(j) - \{i\}} Q_{k,j}^0}{(1 - p_j^0) \prod_{k \in \text{Bit}(j) - \{i\}} (1 - Q_{k,j}^0) + p_j^0 \prod_{k \in \text{Bit}(j) - \{i\}} Q_{k,j}^0}$$

Decision step:

$$P_j^0 = \frac{p_j^0 \prod_{k \in \text{Bit}(j)} Q_{k,j}^0}{(1 - p_j^0) \prod_{k \in \text{Bit}(j)} (1 - Q_{k,j}^0) + p_j^0 \prod_{k \in \text{Bit}(j)} Q_{k,j}^0} \stackrel{1}{\leq} 1 - P_j^0$$

Termination:

If $\hat{\mathbf{c}}\mathbf{H}^T = \mathbf{0}$, the algorithm stops;
else go to Horizontal step.

□ Final remark

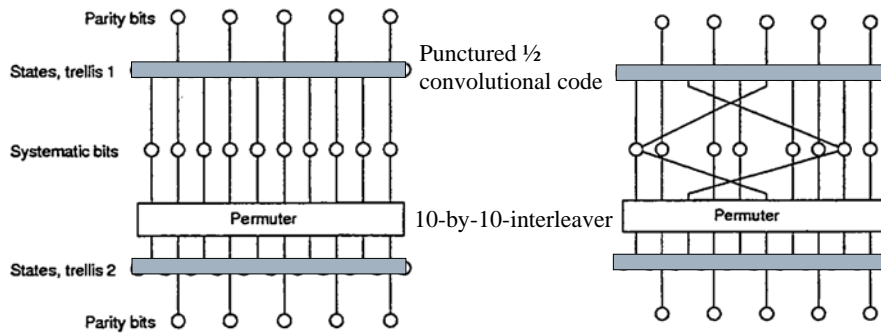
- Regular LDPC codes do not appear to come as close to Shannon's limit as do their turbo code counterparts.
- Hence, irregular LDPC codes are most popular due to this reason.

10.11 Irregular codes

- The performance of turbo codes and LDPC codes can be further improved by “irregularity”.
 - By “irregularity”, we mean that each systematic bit is not used the same number of times.
 - For example, regular turbo code indicates that each systematic bit is used twice in encoding process.

Regular (20, 10) turbo code

Irregular (18, 8) turbo code
(Bits 0 and 6 are used four times,
while bits 1, 2, 3, 4, 5, 7 are used
only twice.)



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Chapter 10-141

10.11 Irregular codes

□ Why irregularity gives better performance?

■ The code word is “bit-wisely dependent”.

```
000000
000111
111000
111111
```

■ If we give a much better estimation on certain positions, e.g., bit 0 and bit 4 in the above example, then the transmitted codeword may be more easily (or earlier-in-iterative-decoding-viewpoint) identified.

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Chapter 10-142

10.11 Irregular codes

