

Chap. 3 Low-Density Parity-Check Codes

□ Review of Block Codes

◇ Block code

- Alphabet \mathcal{A} : finite field \mathbb{F}_q , finite ring \mathbb{Z}_q , etc.
- (n, M) block code \mathcal{C} over \mathcal{A}
= the set of M vectors of length n with components in \mathcal{A}
- Rate = $\frac{\log_q M}{n}$
- A vector $\mathbf{c} \in \mathcal{C}$ is called a codeword or a code vector of \mathcal{C} .

◇ An $[n, k]$ linear block code \mathcal{C} over \mathbb{F}_2

- \mathcal{C} is a k -dimensional subspace of \mathbb{F}_2^n
- $\mathbb{F}_2^n \triangleq \{(x_1, x_2, \dots, x_n) \mid x_i \in \mathbb{F}_2\}$: n -dimensional vector space over \mathbb{F}_2
- $n \triangleq$ length (or code length) of the code
- $k \triangleq$ dimension of the code
- Code rate = $\frac{k}{n}$
- Conditions for a linear code (or a subspace)
 - 1) $\mathbf{c}_1 + \mathbf{c}_2 \in \mathcal{C}, \quad \forall \mathbf{c}_1, \mathbf{c}_2 \in \mathcal{C}$
 - 2) $a\mathbf{c} \in \mathcal{C}, \quad \forall \mathbf{c} \in \mathcal{C}, \quad \forall a \in \mathbb{F}_q$

◇ **Generator matrix G for \mathcal{C}**

- Since \mathcal{C} is a k -dimensional subspace of \mathbb{F}_2^n , there is a basis for \mathcal{C} , say,

$$\{\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_k\}.$$

Note:

- 1) $\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_k$ are linearly independent over \mathbb{F}_q .
 - 2) $\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_k$ span \mathcal{C} , i.e. $\mathcal{C} = \langle \mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_k \rangle$.
- Any codeword $\mathbf{c} = (c_1, c_2, \dots, c_n)$ can be expressed uniquely as a linear combination of $\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_k$. i.e.,

$$\mathbf{c} = m_1 \mathbf{g}_1 + m_2 \mathbf{g}_2 + \dots + m_k \mathbf{g}_k.$$

Therefore,

$$\begin{aligned} \mathbf{c} &= \begin{bmatrix} m_1 & m_2 & \dots & m_k \end{bmatrix} \begin{bmatrix} \mathbf{g}_1 \\ \mathbf{g}_2 \\ \vdots \\ \mathbf{g}_k \end{bmatrix} \\ &= \begin{bmatrix} m_1 & m_2 & \dots & m_k \end{bmatrix} \begin{bmatrix} g_{11} & g_{12} & \dots & g_{1n} \\ g_{21} & g_{22} & \dots & g_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ g_{k1} & g_{k2} & \dots & g_{kn} \end{bmatrix} \end{aligned}$$

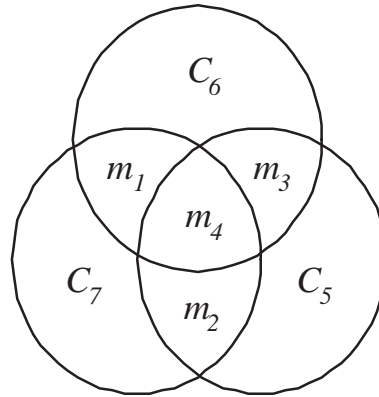
$$\text{or } \mathbf{c} = \mathbf{m}G.$$

- Note:**
- 1) G is called a *generator matrix* of the code \mathcal{C}
 - 2) $\mathcal{C} =$ the row space of $G = \{\mathbf{m}G \mid \mathbf{m} \in \mathbb{F}_2^k\}$
 - 3) $\mathbf{m} \in \mathbb{F}_2^k$: a message to be encoded.

Remark:

- 1) Elementary row operations on G does not change the code \mathcal{C} because $\mathcal{C} = \mathcal{R}(G)$
- 2) Any generator matrix can be reduced to a “*row-reduced echelon form*”
- 3) A linear block code can always be considered to be equivalent to a systematic code by applying elementary column operations, if necessary.
 $\Rightarrow G = [I_k \mid P]$

Example: [7,4] Hamming code



Encoding rule:

$$c_1 = m_1$$

$$c_2 = m_2$$

$$c_3 = m_3$$

$$c_4 = m_4$$

$$c_5 = m_2 + m_3 + m_4 \pmod{2}$$

$$c_6 = m_1 + m_3 + m_4 \pmod{2}$$

$$c_7 = m_1 + m_2 + m_4 \pmod{2}$$

information symbols

redundant symbols

The codeword \mathbf{c} can be expressed as

$$\mathbf{c} = [c_1 \ c_2 \ c_3 \ \cdots \ c_7]$$

$$= [m_1 \ m_2 \ m_3 \ m_4] \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix}$$

$$\Rightarrow \mathbf{c} = \mathbf{m}G$$

On the other hand, \mathbf{c} satisfies the following equations

$$c_2 + c_3 + c_4 + c_5 = 0$$

$$c_1 + c_3 + c_4 + c_6 = 0$$

$$c_1 + c_2 + c_4 + c_7 = 0$$

$$\Rightarrow \begin{bmatrix} 0 & 1 & 1 & 1 & : & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & : & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 & : & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_7 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \quad \text{or} \quad H\mathbf{c}^t = 0$$

- In general,

a matrix H with the property that $H\mathbf{c}^t = 0$ if and only if $\mathbf{c} \in \mathcal{C}$ is called a **parity-check matrix** for \mathcal{C} .

Remark:

- 1) \mathcal{C} is the null space of H , denoted by $\mathcal{C} = \mathcal{N}(H)$, that is,

$$\mathcal{C} = \{\mathbf{c} \in \mathbb{F}_2^n \mid H\mathbf{c}^t = 0\}.$$

- 2) For any $[n, k]$ systematic code,

$$G = [I_k : P] \leftrightarrow H = [-P : I_{n-k}]$$

- 3) $GH^T = 0$

◇ **Minimum distance of \mathcal{C}**

- $w_H(\mathbf{x}) \triangleq$ Hamming weight of \mathbf{x}
= the number of nonzero symbols in \mathbf{x}
- $d_H(\mathbf{x}, \mathbf{y}) \triangleq$ Hamming distance between \mathbf{x} and \mathbf{y}
= $w_H(\mathbf{x} - \mathbf{y})$
- $d_H(\mathbf{x}, \mathbf{y})$ is a metric:
 - a) $d_H(\mathbf{x}, \mathbf{y}) \geq 0$ with equality iff $\mathbf{x} = \mathbf{y}$
 - b) $d_H(\mathbf{x}, \mathbf{y}) = d_H(\mathbf{y}, \mathbf{x})$
 - c) $d_H(\mathbf{x}, \mathbf{y}) \leq d_H(\mathbf{x}, \mathbf{z}) + d_H(\mathbf{z}, \mathbf{y})$
- $d \triangleq$ minimum distance of \mathcal{C}
= $\min\{d_H(\mathbf{x}, \mathbf{y}) \mid \mathbf{x} \neq \mathbf{y}; \mathbf{x}, \mathbf{y} \in \mathcal{C}\}$
= $\min\{w_H(\mathbf{c}) \mid \mathbf{c} \in \mathcal{C} \setminus \{\mathbf{0}\}\}$ (for a linear code)

Theorem 1 *There exists a codeword of weight s in \mathcal{C} iff there exists s columns in H which are linearly dependent.*

Corollary 2 *The minimum distance d of \mathcal{C} is the minimum number of columns in H which are linearly dependent (over \mathbb{F}_q).*

◇ **Repetition code vs. Even parity check code**

$$\left. \begin{aligned} G &= [1 \ 1 \ 1 \ \cdots \ 1] \\ H &= \begin{bmatrix} 1 & 0 & 0 & \cdots & 0 & 1 \\ 0 & 1 & 0 & \cdots & 0 & 1 \\ & & & \vdots & & \\ 0 & 0 & 0 & \cdots & 1 & 1 \end{bmatrix} \end{aligned} \right\} [n, 1, n] \text{ code}$$

◇ **Dual code of \mathcal{C} :**

$$\begin{aligned} \mathcal{C}^\perp &= \text{dual code of } \mathcal{C} \\ &= \{\mathbf{x} \in \mathbb{F}_2^n \mid \mathbf{x} \mathbf{c}^t = 0, \forall \mathbf{c} \in \mathcal{C}\} \end{aligned}$$

□ Low-Density Parity-Check Codes

◇ Milestone References

- 1) R. G. Gallager, "Low-density parity-check codes," *IEEE Trans. Inform. Theory*, pp. 21-28, Jan. 1962.
- 2) R. G. Gallager, *Low-density Parity-check Codes*. MIT Press, 1963.
(An expanded and revised version of Ph.D.thesis (1960))

◇ The Noisy Channel Coding Theorem (Shannon)

If properly coded information is transmitted at a rate R below channel capacity C , then *the probability P_e of decoding error can be made to approach zero exponentially with the code length.*

Note:

1) code length $n \quad \uparrow \quad \Rightarrow \quad$ decoding complexity \uparrow
(computation time or equipment costs)

2) If $R < C$, then

$$P_e < e^{-nE(R)}$$

where $E(R)$ is the error exponent.

◇ Elias (1955)

If *a typical parity-check code of long block length* is used on a binary symmetric channel, and if the code rate is between critical rate and channel capacity, then *the probability of decoding error will be almost as small as that for the best possible code of that rate and block length.*

Note:

In general, decoding of parity-check codes (or linear codes) is not simple.

\Rightarrow Need special classes of parity-check codes.

"Ensemble of LDPC codes"

◇ Low-density parity-check codes

- A low-density parity-check code is a linear block code whose parity check matrix contains most 0's and only a small number of 1's.

“sparseness” or “low-density of 1's”

- A **regular (N, j, k) LDPC code** is a linear block code of length N , whose parity-check matrix H contains exactly j 1's in each column and k 1's in each row.

Note:

- 1) $j, k \ll N$ and $j, k \ll N - K$.
- 2) Assuming that H has full rank, say $N - K$,

$$jN = (N - K)k.$$

Therefore, the *code rate* R is given by

$$R \triangleq \frac{K}{N} = \frac{k - j}{k}$$

where K is the dimension of the code.

- 3) $j \geq 3$ and $k > j$ in most applications.
- 4) For $j > 3$ and a sufficiently low rate, the probability of error over a BSC decreases at least exponentially with \sqrt{N} .

◇ Remark on the Historical Backgrounds

- *Gallager (1960) invented LDPC codes and their iterative decoding.*
- Dark Ages
- Zyablov and Pinsker (1975): Flipping. Linear fraction
- *Tanner (1981): codes defined on graphs*
- *Pearl (1986): Belief propagation*
- Surlas (1989): codes and random fields
- *Berrou, Galvieux, Thitimajshima (1993): turbo codes*

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• *Rediscovery of LDPC codes*

Reinvented - Mackay and Neal (1996)

Decoding on Graphs - Wiberg (1996)

Expander graphs - Spielman, Sipser (1996)

Tornado codes (over the erasure channel)

Luby, Mitzenmacher, Shokrollahi, Spielman, Stemann (1997)

Irregular graphs for error correcting

Luby, Mitzenmacher, Shokrollahi, Spielman (1998)

Note:

- 1) Regular LDPC codes are *not optimum* (in the sense of minimizing probability of decoding error for a given block length).
- 2) The maximum rate at which these codes can be used is bounded below channel capacity.
- 3) A very simple decoding scheme exists.

◇ **Methods of Analysis of Codes**

- Individual codes: difficult if $N \rightarrow \infty$.
- An ensemble of LDPC codes: simpler (by statistical statements)

◇ **Construction of an Ensemble of (N, j, k) LDPC codes**

- 1) H is divided into j submatrices, each containing a single 1 in each column.
- 2) The first of these submatrices contains all its 1's in descending order, i.e., the i th row contains 1's in columns $(i-1)k+1$ to ik .
- 3) The other submatrices are merely column permutations of the first.

Example:

An $(j, 6)$ LDPC code can be constructed by

$$H = \begin{bmatrix} 111111 & & & \\ & 111111 & & \\ & & 111111 & \\ & & & 111111 \\ \dots & \dots & \dots & \dots \\ & \pi_2(H_1) & & \\ \dots & \dots & \dots & \dots \\ & \vdots & & \\ \dots & \dots & \dots & \dots \\ & \pi_j(H_1) & & \end{bmatrix} = \begin{bmatrix} H_1 \\ \dots \\ \pi_2(H_1) \\ \dots \\ \vdots \\ \dots \\ \pi_j(H_1) \end{bmatrix}$$

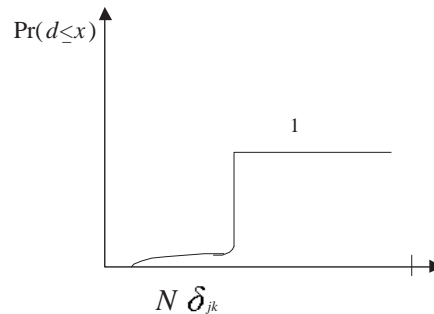
where π_i is a column permutation of H_1

◇ Minimum Distance of LDPC Codes

- The minimum distance d of a code in this ensemble is a *random variable*.
- The distribution function of this random variable can be overbounded by a function.
- For a fixed $j \geq 3$ and $k > j$, as $N \rightarrow \infty$,

$$\Pr(d \leq x) \approx u(x - N\delta_{jk})$$

where δ_{jk} is a fixed fraction and $u(\cdot)$ is the unit-step function.



- For large N , practically all the codes in the ensemble have a minimum distance of at least $N\delta_{jk}$.

| j | k | Rate | δ_{jk} | δ |
|-----|-----|-------|---------------|----------|
| 5 | 6 | 0.167 | 0.255 | 0.263 |
| 4 | 5 | 0.2 | 0.210 | 0.241 |
| 3 | 4 | 0.25 | 0.122 | 0.214 |
| 4 | 6 | 0.333 | 0.129 | 0.173 |
| 3 | 5 | 0.4 | 0.044 | 0.145 |
| 3 | 6 | 0.5 | 0.023 | 0.11 |

δ_{jk} : the ratio of typical minimum distance to block length for an (N, j, k) code
 δ : the same ratio for an ordinary parity-check code of the same rate
 (See Fig 3, Gallager (1962, IT))

- Loss of rate associated with regular LDPC Codes

| j | k | Rate for (N, j, k) code | Rate for an equivalent optimum code of the same exponent |
|-----|-----|--------------------------------|---|
| 3 | 6 | 0.5 | 0.555 |
| 3 | 5 | 0.4 | 0.43 |
| 4 | 6 | 0.333 | 0.343 |
| 3 | 4 | 0.25 | 0.266 |

(See Fig 4, Gallager (1962, IT))

- Over a reasonable range of channel transition probabilities, the low-density code has a probability of decoding error that decreases exponentially with block length and the exponent is the same as that for the optimum code of slightly higher rate.

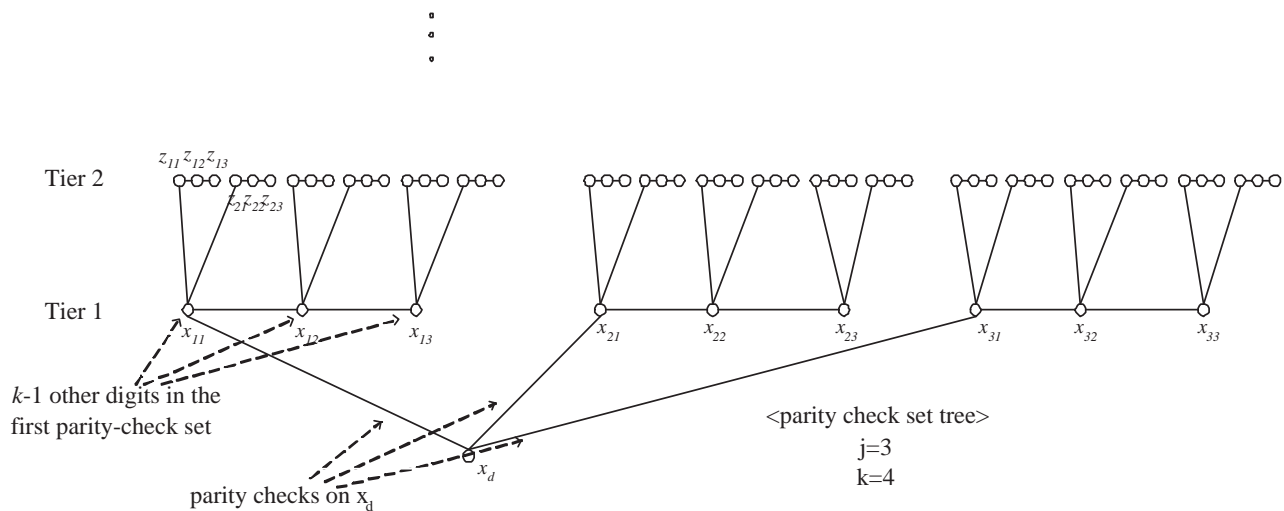
Remark: For long binary BCH codes,

$$\frac{d}{n} \rightarrow 0 \quad \text{as} \quad n \rightarrow \infty \quad (\text{Berlekamp})$$

□ Hard-decision decoding (Gallager Decoding A)

- BSC at rates far below channel capacity
- Decoding procedure:
 - 1) Compute all the parity checks.
 - 2) Change any digit that is contained in more than some fixed number of unsatisfied parity-check equations
 - 3) Recompute the parity checks using these new values
 - 4) Repeat 2) and 3) until the parity checks are all satisfied.

- *Parity-check set tree* ($j = 3$ and $k = 4$ case)



- Parity-check equations on x_d :

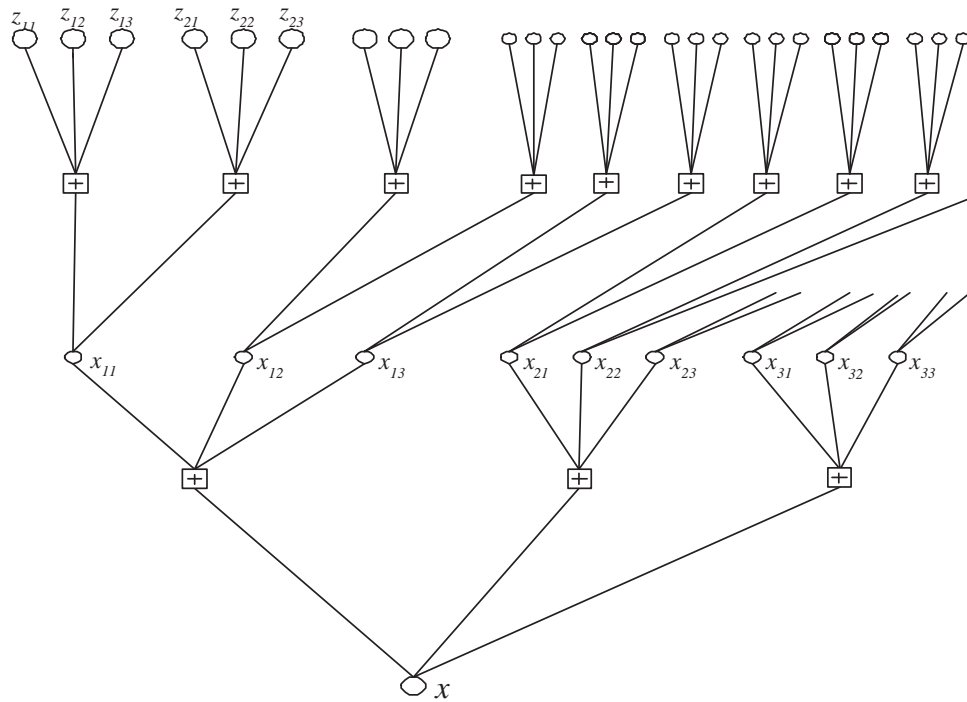
$$\begin{cases} x_d + x_{11} + x_{12} + x_{13} = 0 \\ x_d + x_{21} + x_{22} + x_{23} = 0 \\ x_d + x_{31} + x_{32} + x_{33} = 0 \end{cases}$$

- Parity-check equations on x_{11} :

$$\begin{cases} x_{11} + z_{11} + z_{12} + z_{13} = 0 \\ x_{11} + z_{21} + z_{22} + z_{23} = 0 \\ x_{11} + x_{12} + x_{13} + x_d = 0 \end{cases}$$

◇ **Factor graph**

The factor graph shows the tree structure from node x .



□ Probabilistic Decoding (Belief-Propagation Decoding)

◇ Problem Formulation

- Define

$\{y\} \triangleq$ the set of received symbols

$S \triangleq$ the event that the transmitted digits satisfy the j parity-check equations on digit d .

- Assume the *Ensemble of events* in which

- 1) the transmitted digits in the positions of the first tier of digit d are independent equiprobable binary digits; and
- 2) the probabilities of the received symbols in these positions are determined by the channel transition probabilities $P_x(y)$.

- **Goal:** Compute $\Pr(x_d = 1|\{y\}, S)$.

Theorem 3 Let

$$P_d \triangleq \Pr(x_d = 1|y_d)$$

$$P_{il} \triangleq \Pr(x_{il} = 1|y_{il})$$

where

$x_{il} = l^{\text{th}}$ digit in the i^{th} parity-check set of the first tier.

Then

$$\frac{\Pr(x_d = 0|\{y\}, S)}{\Pr(x_d = 1|\{y\}, S)} = \frac{1 - P_d}{P_d} \prod_{i=1}^j \left[\frac{1 + \prod_{l=1}^{k-1} (1 - 2P_{il})}{1 - \prod_{l=1}^{k-1} (1 - 2P_{il})} \right].$$

Lemma 4 (Sum of I.I.D. Random Variables) *Let z be the sum of independent and identically distributed (i.i.d.) random variables, given by*

$$z = z_1 + z_2 + \cdots + z_m,$$

where z_i 's are independent random variable taking on 0 and 1 with $\Pr(z_i = 1) \triangleq p_i$. Then

$$\Pr(z = \text{even}) = \frac{1 + \prod_{i=1}^m (1 - 2p_i)}{2}.$$

Proof. Note that

$$\begin{aligned} \prod_{i=1}^m [(1 - p_i) + p_i t] &= \sum_{l=0}^m A_l t^l, \\ \prod_{i=1}^m [(1 - p_i) - p_i t] &= \sum_{l=0}^m B_l t^l \end{aligned}$$

where

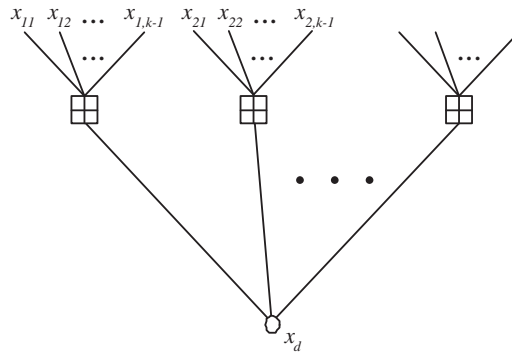
$$A_l = \Pr(z = l) \quad \text{and} \quad B_l = \begin{cases} \Pr(z = l), & l = \text{even} \\ -\Pr(z = l), & l = \text{odd}. \end{cases}$$

Therefore,

$$\begin{aligned} \Pr(z = \text{even}) &= \sum_{l=0, l=\text{even}}^m \Pr(z = l) \\ &= \frac{1}{2} \left[\prod_{i=1}^m (1 - p_i + p_i t) + \prod_{i=1}^m (1 - p_i - p_i t) \right]_{t=1} \\ &= \frac{1 + \prod_{i=1}^m (1 - 2p_i)}{2}. \end{aligned}$$

□

Proof of Theorem:



Assuming that x_{ij} 's are statistically independent,

$$\begin{aligned}
 \Pr(S | x_d = 0, \{y\}) &= \Pr\left(\sum_{l=1}^{k-1} x_{il} = \text{even}, \forall i = 1, \dots, j | x_d = 0, \{y\}\right) \\
 &= \prod_{i=1}^j \left[\frac{1 + \prod_{l=1}^{k-1} (1 - 2P_{il})}{2} \right] \\
 \Pr(S | x_d = 1, \{y\}) &= \Pr\left(\sum_{l=1}^{k-1} x_{il} = \text{odd}, \forall i = 1, \dots, j | x_d = 1, \{y\}\right) \\
 &= \prod_{i=1}^j \left[\frac{1 - \prod_{l=1}^{k-1} (1 - 2P_{il})}{2} \right]
 \end{aligned}$$

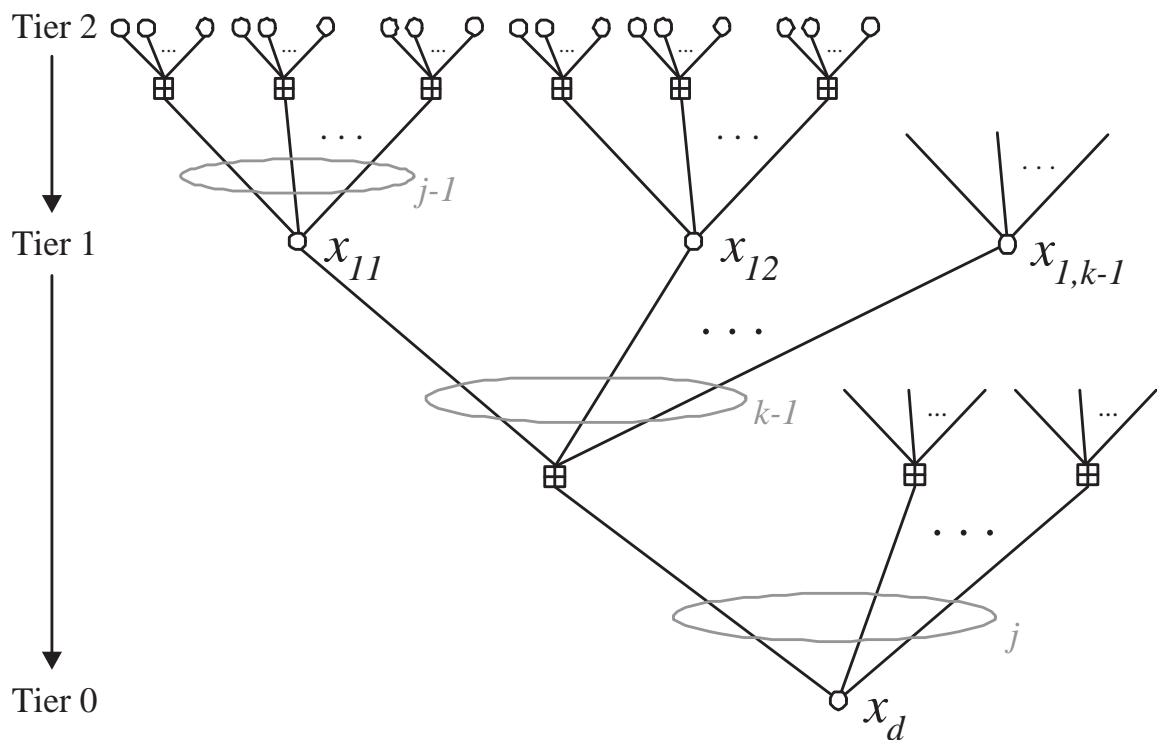
by Lemma 4.

Therefore,

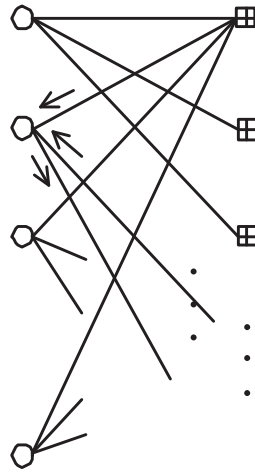
$$\begin{aligned}
 \frac{\Pr(x_d = 0 | \{y\}, S)}{\Pr(x_d = 1 | \{y\}, S)} &= \frac{1 - P_d}{P_d} \cdot \frac{\Pr(S | x_d = 0, \{y\})}{\Pr(S | x_d = 1, \{y\})} \\
 &= \frac{1 - P_d}{P_d} \prod_{i=1}^j \left[\frac{1 + \prod_{l=1}^{k-1} (1 - 2P_{il})}{1 - \prod_{l=1}^{k-1} (1 - 2P_{il})} \right].
 \end{aligned}$$

Note:

- The complexity of decoding in many-tier case can be solved from the 1-tier case by a simple *iterative technique*.
- 2-tier case



□ General Decoding Procedure



Remarks on Gallager Decoding

- 1) If the decoding is successful, $P_d \rightarrow 0$ or 1 (depending on the transmitted digit) as the number of iterations is increased.
- 2) This procedure is only valid for as many iterations as meet the independence assumption.
- 3) cycle \leftrightarrow independency.
- 4) **Reasonable assumption**: The dependencies have a relatively minor effect and tend to cancel each other out somewhat.
- 5) The computation per digit per iteration is independent of block length.
- 6) The average number of iterations required to decode is bounded by a quantity proportional to the $\log(\log n)$.

□ Decoding in terms of LLR

Define

$$\ln \frac{1 - P_d}{P_d} \triangleq \alpha_d \beta_d, \quad \ln \frac{1 - P_{il}}{P_{il}} \triangleq \alpha_{il} \beta_{il}$$

and

$$\ln \frac{\Pr(x_d = 0 | \{y\}, S)}{\Pr(x_d = 1 | \{y\}, S)} \triangleq \alpha'_d \beta'_d$$

($\alpha = \text{sign}$ and $\beta = \text{magnitude}$)

Then

$$\alpha'_d \beta'_d = \alpha_d \beta_d + \sum_{i=1}^j \left(\prod_{l=1}^{k-1} \alpha_{il} \right) f \left(\sum_{l=1}^{k-1} f(\beta_{il}) \right) \quad (1)$$

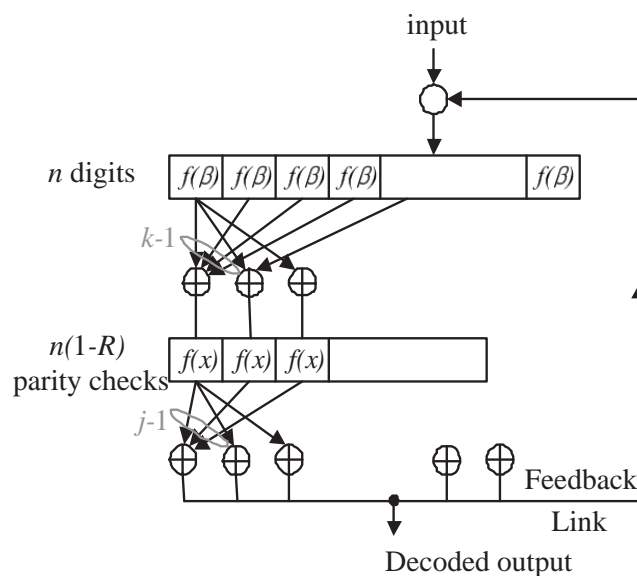
where

$$f(\beta) = \ln \frac{e^\beta + 1}{e^\beta - 1}$$

Exercise: Derive (1).

Note:

- Serial computing
- Parallel computing → fast



Derivation of Equation (1):

$$\begin{aligned}
\ln \frac{1 - P_{il}}{P_{il}} = \alpha_{il} \beta_{il} &\Rightarrow 1 - 2P_{il} = \frac{e^{\alpha_{il} \beta_{il}} - 1}{e^{\alpha_{il} \beta_{il}} + 1} \\
&= \alpha_{il} \frac{e^{\beta_{il}} - 1}{e^{\beta_{il}} + 1} \\
&= \alpha_{il} e^{-f(\beta_{il})}.
\end{aligned}$$

Using this representation, we get

$$\begin{aligned}
\prod_{l=1}^{k-1} (1 - 2P_{il}) &= \left(\prod_{l=1}^{k-1} \alpha_{il} \right) e^{-\sum_{l=1}^{k-1} f(\beta_{il})} \\
&\triangleq \alpha e^{-\beta}.
\end{aligned}$$

where

$$\alpha = \prod_{l=1}^{k-1} \alpha_{il}, \quad \beta = \sum_{l=1}^{k-1} f(\beta_{il}).$$

Hence,

$$\begin{aligned}
\ln \frac{1 + \prod_{l=1}^{k-1} (1 - 2P_{il})}{1 - \prod_{l=1}^{k-1} (1 - 2P_{il})} &= \ln \frac{1 + \alpha e^{-\beta}}{1 - \alpha e^{-\beta}} \\
&= \alpha \ln \frac{1 + e^{-\beta}}{1 - e^{-\beta}} \\
&= \alpha f(\beta).
\end{aligned}$$

$$\therefore \alpha'_d \beta'_d = \alpha_d \beta_d + \sum_{i=1}^j \left(\prod_{l=1}^{k-1} \alpha_{il} \right) f \left(\sum_{l=1}^{k-1} f(\beta_{il}) \right).$$

□ Probability of Error in Gallager Decoding A

Assumption

- 1) BSC with crossover probability p_0 .
- 2) (n, j, k) LDPC code with $j = 3$, for simplicity.

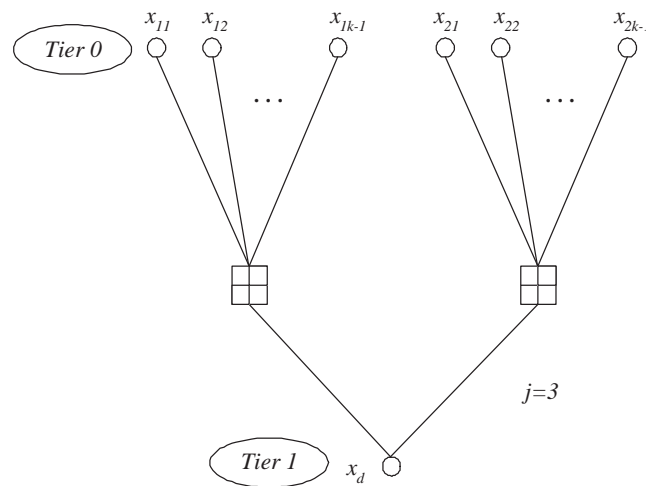
Decoding rule

If *both of two* check-equations containing x_d are not satisfied, say,

$$x_d + x_{11} + \cdots + x_{1,k-1} \neq 0$$

$$x_d + x_{21} + \cdots + x_{2,k-1} \neq 0$$

then $x_d \rightarrow x_d + 1$.



Computation of related probabilities

- $\Pr\{\text{even number of errors among } k - 1 \text{ digits}\}$

$$= \frac{1 + (1 - 2p_0)^{k-1}}{2}.$$

- $\Pr\{\text{odd number of errors among } k - 1 \text{ digits}\}$

$$= \frac{1 - (1 - 2p_0)^{k-1}}{2}.$$

- $\Pr\{\text{a digit in the first tier is received in error and then corrected}\}$

$$= p_0 \cdot \left[\frac{1 + (1 - 2p_0)^{k-1}}{2} \right]^2.$$

- $\Pr\{\text{a digit in the first tier is received correctly but then changed because of unsatisfied parity checks}\}$

$$= (1 - p_0) \cdot \left[\frac{1 - (1 - 2p_0)^{k-1}}{2} \right]^2.$$

- Let p_1 be the probability of error of a digit in the first tier after applying this decoding process.

Then

$$\begin{aligned} p_1 &= p_0 \cdot \Pr\{\text{no correction}\} + (1 - p_0) \cdot \Pr\{\text{correction}\} \\ &= p_0 \left(1 - \left[\frac{1 + (1 - 2p_0)^{k-1}}{2} \right]^2 \right) + (1 - p_0) \left[\frac{1 - (1 - 2p_0)^{k-1}}{2} \right]^2. \end{aligned}$$

- Using the induction, we get the following recursion:

$$p_{i+1} = p_0 \left(1 - \left[\frac{1 + (1 - 2p_i)^{k-1}}{2} \right]^2 \right) + (1 - p_0) \left[\frac{1 - (1 - 2p_i)^{k-1}}{2} \right]^2$$

where p_i is *the probability of error after processing of a digit in the i th tier*.

Decoding convergence

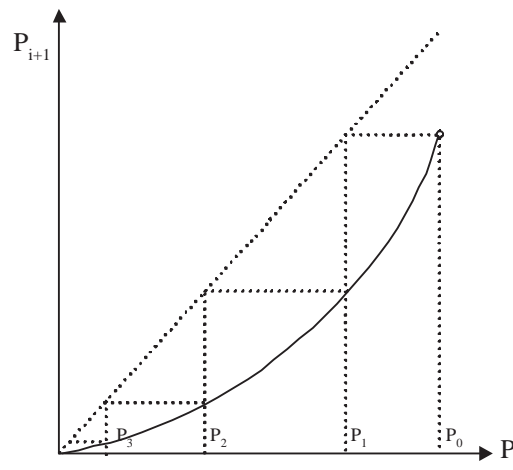
- For sufficiently small p_0 ,

$$p_i \rightarrow 0 \quad \text{as } i \rightarrow \infty.$$

- Note that

$$\left. \begin{array}{ll} 0 < p_{i+1} < p_i & \text{for } 0 < p_i \leq p_0 \\ p_{i+1} = p_i & \text{for } p_i = 0 \end{array} \right\} \Rightarrow p_i \rightarrow 0.$$

- Density evolution from p_i to p_{i+1} :



For sufficiently small p_0 , the recursion for p_i can be approximated as

$$p_{i+1} \approx p_i \cdot 2(k-1)p_0.$$

Therefore,

$$p_i \approx c [2(k-1)p_0]^i$$

for a constant c .

Maximum p_0 for weak bound decoding convergence in regular LDPC codes

| j | k | Rate | p_0 |
|-----|-----|-------|-------|
| 3 | 6 | 0.5 | 0.04 |
| 3 | 5 | 0.4 | 0.061 |
| 4 | 6 | 0.333 | 0.075 |
| 3 | 4 | 0.25 | 0.106 |

□ General Case

Decoding rule:

$$x \rightarrow x + 1 \pmod{2} \quad \text{if the number of unsatisfied parity-checks} \geq b.$$

Recursion for p_i :

$$\begin{aligned} p_{i+1} = & p_0 - p_0 \sum_{l=b}^{j-1} \binom{j-1}{l} \left[\frac{1 + (1 - 2p_i)^{k-1}}{2} \right]^l \left[\frac{1 - (1 - 2p_i)^{k-1}}{2} \right]^{j-1-l} \\ & + (1 - p_0) \sum_{l=b}^{j-1} \binom{j-1}{l} \left[\frac{1 - (1 - 2p_i)^{k-1}}{2} \right]^l \left[\frac{1 + (1 - 2p_i)^{k-1}}{2} \right]^{j-1-l}. \end{aligned}$$

Convergence of p_i :

- To minimize p_i , find the smallest integers b for which

$$\frac{1 - p_0}{p_0} \leq \left[\frac{1 + (1 - 2p_i)^{k-1}}{1 - (1 - 2p_i)^{k-1}} \right]^{2b-j+1}.$$

Note that $p_i \downarrow \Rightarrow b \downarrow$

- Optimal choice for b , if p_i is sufficiently small:

$$b = \begin{cases} j/2, & j \text{ even} \\ (j+1)/2, & j \text{ odd.} \end{cases}$$

- Then the recursion for p_i can be approximated as

$$p_{i+1} = \begin{cases} p_0 \binom{j-1}{\frac{j-1}{2}} (k-1)^{(j-1)/2} p_i^{(j-1)/2} + \text{higher order terms} & (j \text{ odd}) \\ \binom{j-1}{\frac{j}{2}} (k-1)^{j/2} p_i^{j/2} + \text{higher order terms} & (j \text{ even}). \end{cases}$$

Therefore, p_i can be approximated as

$$p_i \leq \begin{cases} \exp \left[-C_{jk} \left(\frac{j-1}{2} \right)^i \right], & j \text{ odd} \\ \exp \left[-C_{jk} \left(\frac{j}{2} \right)^i \right], & j \text{ even.} \end{cases}$$

Number of iterations to guarantee the independency:

- Since there are $(j-1)^m(k-1)^m$ digits in the m th tier of a tree,

$$n \geq (j-1)^m(k-1)^m$$

for independent digits.

- Then

$$\frac{\ln n}{\ln(j-1)(k-1)} \geq m \geq \frac{\ln \left(\frac{n}{2k} - \frac{n}{2j(k-1)} \right)}{2 \ln(k-1)(j-1)}.$$

- The probability p_m of error after processing of a digit in the m th tier can be upper bounded by

$$p_m \leq \exp \left(-c_{jk} \left[\frac{n}{2k} - \frac{n}{2j(k-1)} \right]^\alpha \right)$$

where

$$\alpha = \begin{cases} \frac{\ln \frac{j-1}{2}}{2 \ln(k-1)(j-1)}, & j \text{ odd} \\ \frac{\ln \frac{j}{2}}{2 \ln(k-1)(j-1)}, & j \text{ even.} \end{cases}$$

- For $j > 3$,

$$p_m \leq \exp(-c\sqrt{n}).$$

Remark:

- 1) Another way to evaluate the probabilistic decoding scheme is to *calculate the probability distributions of the log-likelihood ratios (LLRs) for a number of iterations*.
- 2) The above approach is also another *density evolution*.