

Lecture 10 Constraints & LDPC Codes February 13, 2024

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Announcements & Agenda

Announcements

Today

- Midterm Review
- Turbo-Code Completion
- Soft Information from constraints
- LDPC Codes
- Hard/Soft concatenation Reed Solomon outer

Problem Set 5 = PS5 due Wednesday February 21

- 1. 8.12 Turbo Design and Coding
- 2. 8.13 Constraints and BICM
- 3. 8.14 LDPC Use
- 4. 8.15 Subsymbol- vs Symbol-Level Deterministic Interleaving
- 5. 8.16 Wireless Hard-Soft Interleaving Challenge





Solutions

2.



- 1. QAM Design
 - a. Attenuation -63-(-40-70) = <mark>47 dB</mark>; SNR =-110-(-142+3) = <mark>29dB</mark>

b.
$$P_e = 4\left(1 - \frac{1}{16}\right) Q\left(\sqrt{\frac{3 \cdot 10^{2.9}}{255}}\right) = .0042$$

- *c.* $C = 10^7 \cdot log_2(1+10^{2.9}) = 96Mbps$
- d. SNRnew = 10*log10(255/3*(qfuncinv(1e-6/3.75))^2) = 33.3 +.2, γ = 4.5 or 4.8 dB
 a. dfree=6 and G=[17 13] from tables.
- e. $SNR_{min} = 0$ ($-\infty dB$) and $SNR_{max} = +\infty dB$ (also +26 and smaller)

$$f. \quad \langle \mathbf{P}_{e} \rangle = \frac{1}{2} \left[1 - \sqrt{\frac{\kappa SNR}{\kappa SNR+1}} \right] = .0248$$

- g. SNRnew = 10*log10(63/3*(qfuncinv(1e-6/3.5))^2) = 27.3 dB
 - This is 1.7 dB below average at 29 dB ; $10^{-1.7} = .6761$

Coding of course will help.

$$P_{out} = \int_0^{.6761} \frac{1}{2} \cdot e^{-x/2} \cdot dx = 1 - e^{-.6761/2} = .2868$$

$$\begin{array}{rcl} g & \geq & -1.7 - 6.3 = -8 \ \mathrm{dB} = .1585 \\ P_{out} & = & .1 - e^{-.1585/2} = 0762 \ \mathrm{or} \ 7.6\% \\ g & \geq & -14.6 \ \mathrm{dB}(.0316) \\ P_{out} & = & 1 - e^{10^{-.0316/2}} = .0157 \ \mathrm{or} \ 1.6 \ \% \end{array}$$

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Bridge SOVA / APP *a.* $2^6 = 64$ paths; $2^{\nu} = 4$ survivors 2⁶ x 2¹² table entries *b.* $\frac{p_{v'=00}}{v} = \begin{cases} (1-p)^2 & v = 00\\ (1-p) \cdot p & v = 01\\ p \cdot (1-p) & v = 10\\ p^2 & v = 11 \end{cases}$ C. No, encoder has memory d. 4 survivors е. $p_{\boldsymbol{u}=000011/\boldsymbol{v}'} = p^2 \cdot (1-p)^{10}$ $p_{u=110110/v'} = p^3 \cdot (1-p)^9$ $p_{u=110101/v'} = p^4 \cdot (1-p)^8$ $p_{u=110100/v'} = p^4 \cdot (1-p)^8$ Non-survivors' prob f. $1-p^2 \cdot (1-p)^{10} - p^3 \cdot (1-p)^9 - 2p^4 \cdot (1-p)^8$ g. $LLR_3 = ln(\frac{7}{3})$ $u_3 = 0$: 1 branch $1 \cdot \left\{ p^2 \cdot (1-p)^{10} \right\} = .0035$ $u_3 = 1$: 3 branches $2 \cdot \{p^4 \cdot (1-p)^8\} + \{p^3 \cdot (1-p)^9\} = .0020$ h. Less confidence than APP, better than SOVA

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Soft Information from Constraints

Section 7.4

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Block Codes aggregate many tiny codes

Each row in an arbitrary linear encoder's parity matrix can be viewed as a simple linear parity code.

- Constraints
 - $p_{x/y}$ is the probability of x given that the *decision* based on y meets code/modulation constraints.
- $p_{x/constraints} \propto p_{x,constraints} = p_{intrinsic} \cdot p_{extrinsic}$

current x's other x's

- For instance, the parity-check equation $v \cdot H = 0$ provides n k parity constraints.
- So p_{x/constraints} essentially means the MAP finds the x most likely to satisfy all these parity constraints.
- There are other types of constraints also:
 - Equality constraints These basically recognize that any x (often a bit) must have common decision in every constraint in which it participates.
 - Modulator constraints Only certain constellation-to-bit mappings may occur for particular modulator (BICM).
 - **Channel-Model constraints** Certain bit/x combinations may not be likely, given a certain channel filter.
 - These sometimes have the name "turbo equalization."



Section 7.4

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Basics PRIOR to the constraint

- BSC for *i* = 1,2,9
 - Before constraint:

$$p(v_{i}, y_{i}) = p(y_{i}/v_{i}) \cdot p(v_{i})$$

$$p(v_{i}, y_{i}) = \begin{cases} (1-p) \cdot (p_{i}) & y_{i} = 1, v_{i} = 1 \\ p \cdot (1-p_{i}) & y_{i} = 1, v_{i} = 0 \\ p \cdot p_{i} & y_{i} = 0, v_{i} = 1 \\ (1-p) \cdot (1-p_{i}) & y_{i} = 0, v_{i} = 0 \end{cases}$$
intrinsic Extrinsic Info on *i* from *j* \neq *i*

- AWGN for *i* = 1,2,9
 - Before constraint:

$$p(v_i, y) = \begin{cases} \underbrace{\frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{-\frac{1}{2\sigma^2}(y - \sqrt{\bar{\mathcal{E}}} x)^2}}_{p_{int}} \cdot \underbrace{\frac{p_i}{p_{ext}}}_{p_{ext}} & v_i = 1\\ \underbrace{\frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{-\frac{1}{2\sigma^2}(y + \sqrt{\bar{\mathcal{E}}} x)^2}}_{p_{int}} \cdot \underbrace{\frac{p_i}{p_{ext}}}_{p_{ext}} & v_i = 0 \end{cases}$$



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Section 7.4.1

Example parity constraint

- Example has 3 bits in a specific **parity equation** (row of H, call it h_i , or column of H^t); $H \rightarrow G$.
 - **Generator** G's output is such that $v_1 \oplus v_2 \oplus v_9 = 0$; this corresponds to 1's in positions 1,2, and 9 in a row of H.
 - First: A BSC with bit-error parameter p has channel outputs y_1 , y_2 , y_9 and encoder outs v_1 , v_2 , v_9 .
- S_E is a subset $S_E = \{ \boldsymbol{v} \mid E(\boldsymbol{v}) = 0 \}$ all the bit combinations that satisfy the constraint: $S_E = \{(0,0,0), (1,1,0), (1,0,1), (0,1,1)\}$
- $S_{E\setminus i}(y_i)$ fixes each set-codeword's position *i* to be the specific value y_i .
 - $S_{E\setminus3}(y_3 = \mathbf{0}) = \{(\mathbf{0}, 0, 0), (\mathbf{0}, 1, 1)\}$
- MAP decoder to $v_{i=3}$ for this event satisfies $\max_{v_3 \in \{0,1\}} p_{v_{3/E}}$



$$p_{v_3/E} \propto p_{v_3,E} = p_{ext}(v_3/E, y_3) \cdot p_{int}(E, y_3)$$

$$p_{ext}(v_3/E, y_3) \propto \begin{cases} \Pr\{v_3 = y_3 = 0\} = p_1 \cdot p_2 + (1 - p_1) \cdot (1 - p_2) \\ \Pr\{v_3 = y_3 = 1\} = p_1 \cdot (1 - p_2) + (1 - p_1) \cdot p_2 \end{cases}$$

$$p_{int}(E, y_3) \propto \begin{cases} p_3 & ; y_3 = 1 \\ 1 - p_3 & ; y_3 = 0 \end{cases} \qquad p_3 = p_{BSC} = p$$
L10:7 Stanford University

Section 7.4.1

MAP Decoder maximizes $p_{v_i/E}$

• For bit
$$i = 3$$
:

$$p_{v_3,E} = \frac{1}{c'_3} \cdot \begin{cases} p_1 \cdot p_2 \cdot (1-p_3) + (1-p_1) \cdot (1-p_2) \cdot (1-p_3) & v_3 = 0 \\ p_1 \cdot (1-p_2) \cdot p_3 + (1-p_1) \cdot p_2 \cdot p_3 & v_3 = 1 \end{cases}$$

• For bit
$$i = 2$$
:

$$p_{v_2,E} = \frac{1}{c'_2} \cdot \begin{cases} p_1 \cdot p_3 \cdot (1-p_3) + (1-p_1) \cdot (1-p_3) \cdot (1-p_2) & v_2 = 0 \\ p_1 \cdot (1-p_3) \cdot p_2 + (1-p_1) \cdot p_2 \cdot p_3 & v_2 = 1 \end{cases}$$

• For bit
$$i = 1$$
:

$$p_{v_1,E} \frac{1}{c'_1} \cdot \begin{cases} p_3 \cdot p_2 \cdot (1-p_1) + (1-p_3) \cdot (1-p_2) \cdot (1-p_1) & v_1 = 0 \\ p_3 \cdot (1-p_2) \cdot p_1 + (1-p_3) \cdot p_2 \cdot p_1 & v_1 = 1 \end{cases}$$



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Events and their probability calculation

- Satisfaction of parity check is an example of, more generally, an event E(v) = 0.
- S_E is a subset $S_E = \{ v \mid E(v) = 0 \}$ all the bit combinations that satisfy the constraint.
 - $S_{E\setminus i}(y_i)$ fixes each set-codeword's position *i* to be the specific value y_i .
- MAP decoder for this event satisfies $\max_{v_i \in \{0,1\}} p_{v_{i/E}}$. $p_{ext}(y_i / E, y_i) = c_i \cdot \sum_{\substack{v \in S_{E \setminus i}(y_i) \\ j \neq i}} \prod_{\substack{j=1 \\ j \neq i}}^n p_j(E, y_i)$ Recall L10:7 **BSC/AWGN** $c_i = \left\{ \sum_{y \in S_n} \prod_{i=1}^n p_i(E, y_i) \right\}^{-1}$ $p_{ext}(v_3 / E, y_3) \propto \begin{cases} \Pr\{v_3 = y_3 = 0\} = p_1 \cdot p_2 + (1 - p_1) \cdot (1 - p_2) \\ \Pr\{v_3 = y_3 = 1\} = p_1 \cdot (1 - p_2) + (1 - p_1) \cdot p_2 \end{cases}$ $c_3 = \frac{1}{1 - 2 \cdot n_1 \cdot n_2}$ Similarly, for v_1 and v_2 - send p_{ext} to 3 other constraint decoders



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Section 7.4.1

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Soft Bits

• The soft bit is $\chi_i = 2 \cdot Pr\{v_i = 0\} - 1 = 1 - 2 \cdot Pr\{v_i = 1\}.$

A soft bit accepts any probability (extrinsic , intrinsic, ...) for Pr{v_i = 0}.

- The soft bit relates to LLR as $LLR_i = \ln \frac{\chi_i + 1}{\chi_i 1}$ or $\chi_i = -tanh\left(\frac{LLR_i}{2}\right)$.
- By induction (with t_r = # of 1's in a row)
- Use this soft bit with extrinsic information for all the "other" bits:

Define the involution
$$\phi(x) = \phi^{-1}(x) = -\ln\left[\tanh\left(\frac{x}{2}\right)\right] = \ln\left(\frac{e^x + 1}{e^x - 1}\right)$$

 $\chi_i = \prod_{j=1}^{c_r} \chi_j \, .$

• So then
$$\phi(LLR_{ext,i}) \stackrel{\Delta}{=} + \ln\left(\frac{e^{LLR_{ext,i}} + 1}{e^{LLR_{ext,i}} - 1}\right) = -\ln\left(\tanh\left[\frac{LLR_{ext,i}}{2}\right]\right)$$

- And finally: $\chi_i \cdot \chi_j \leftrightarrow \phi(LLR_i) + \phi(LLR_j)$.
 - This means no multiplication, just adds and table look-up $\phi(x)$.

Illustration on next slide



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Section 7.4.1.1

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Parity Constraint Soft-Information Flows



• So each bit, considered like a tiny code, sends receives extrinsic info and sends intrinsic info, to all others in *E*.



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Section 7.4.1.1

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Equality Constraints

• Each bit may participate in many constraints – it should ultimately have same value in them all.



$$S_E = \{(0,0,0), (1,1,1)\}$$

$$p_{ext}(v_2 / E, y_2) = c_2 \cdot \begin{cases} a_1 \cdot a_3 & v_i(2) = 1\\ (1 - a_1) \cdot (1 - a_3) & v_i(2) = 0 \end{cases}$$
$$c_2 = \frac{1}{a_1 \cdot a_3 + (1 - a_1) \cdot (1 - a_3)}$$

$$p_{int}(E, y_2) \propto \begin{cases} p_2 & ; y_2 = 1\\ 1 - p_2 & ; y_2 = 0 \end{cases}$$

Equality-Constraint Decoder maximizes

$$p_{v_i,E} = c'_i \cdot \begin{cases} a_1 \cdot a_2 \cdot a_3 & v_i = 1\\ (1 - a_1) \cdot (1 - a_2) \cdot (1 - a_3) & v_i = 0 \end{cases}$$
$$c'_i = \frac{1}{a_1 \cdot a_2 \cdot a_3 + (1 - a_1) \cdot (1 - a_2) \cdot (1 - a_3)}$$



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Section 7.4.2

Equality Constraint Soft-Information Flows

 The extrinsic information returns to other (e.g., parity) constraints, and the constraint accepts intrinsic from others



- The Equality and Parity constraints for a binary block code can thus cycle soft information.
- This is another form of iterative decoding.



Section 7.4.2.1

Simple Iterative Decoder Illustration

- It's called a "Tanner Graph" or "Factor Graph."
- Decoding may take multiple iterations:
 - When extrinsic data from an equality node cycles back to that same node, the soft-information can become "biased."
 - Such a biased decoder then loses exact MAP quality.
- Good codes try to make the cycle longer than the number of iterations that lead to convergence.
 - This can only be done approximately in practice.
- Good LDPC codes achieve this.
 - Designers actually design the *H* matrix .
 - And then just use a corresponding systematic G.
 - Do this by simple row add operations to designed H to $H_{sys} = [h \ I]$.
 - $G = \begin{bmatrix} I & h^t \end{bmatrix}$ so then $G \cdot H^t = G \cdot H^t_{sys} = 0$.





Section 7.5.3

Soft-Information from constellation



Basically, sum contributions for common v_i values of 0 and then 1, normalizing the constant as follows:

$$\begin{split} p(y_1 &= -5.5, v_3 = 0) &= c_1 \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \left(e^{-\frac{1}{2\sigma^2}(.5)^2} + e^{-\frac{1}{2\sigma^2}(1.5)^2} + e^{-\frac{1}{2\sigma^2}(2.5)^2} + e^{-\frac{1}{2\sigma^2}(4.5)^2} \right) \cdot (1 - p_3) \\ p(y_1 &= -5.5, v_3 = 1) &= c_1 \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \left(e^{-\frac{1}{2\sigma^2}(6.5)^2} + e^{-\frac{1}{2\sigma^2}(8.5)^2} + e^{-\frac{1}{2\sigma^2}(10.5)^2} + e^{-\frac{1}{2\sigma^2}(12.5)^2} \right) \cdot p_3 \quad , \\ p(y_1 &= -5.5, v_2 = 0) &= c_2 \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \left(e^{-\frac{1}{2\sigma^2}(.5)^2} + e^{-\frac{1}{2\sigma^2}(1.5)^2} + e^{-\frac{1}{2\sigma^2}(10.5)^2} + e^{-\frac{1}{2\sigma^2}(12.5)^2} \right) \cdot (1 - p_2) \\ p(y_1 &= -5.5, v_2 = 1) &= c_2 \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \left(e^{-\frac{1}{2\sigma^2}(2.5)^2} + e^{-\frac{1}{2\sigma^2}(4.5)^2} + e^{-\frac{1}{2\sigma^2}(6.5)^2} + e^{-\frac{1}{2\sigma^2}(8.5)^2} \right) \cdot p_2 \quad , \\ p(y_1 &= -5.5, v_1 = 0) &= c_3 \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \left(e^{-\frac{1}{2\sigma^2}(1.5)^2} + e^{-\frac{1}{2\sigma^2}(2.5)^2} + e^{-\frac{1}{2\sigma^2}(8.5)^2} + e^{-\frac{1}{2\sigma^2}(10.5)^2} \right) \cdot (1 - p_1) \\ p(y_1 &= -5.5, v_1 = 1) &= c_3 \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \left(e^{-\frac{1}{2\sigma^2}(.5)^2} + e^{-\frac{1}{2\sigma^2}(4.5)^2} + e^{-\frac{1}{2\sigma^2}(6.5)^2} + e^{-\frac{1}{2\sigma^2}(10.5)^2} \right) \cdot p_1 \quad . \end{split}$$



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Section 7.4.3

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LDPC Codes

Section 8.3.3

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LDPC as "almost random" codes

- R. Gallager (MIT), early 1960's, designed the parity check matrix *H* directly (e.g., design the null space / checks).
 - His code ensemble averaged n k parity bits that were randomly placed in H for given large n.
 - r = k/n is finite.
 - The "low density" (LD) part → SPARSE BINARY MATRIX (see matlab's "sparse.m" and "nnz.m" commands).
- The ensemble works at capacity limit; RG even found some codes that were really good.
- However, the consequent ML Decoders however were too complex!
- 1990's Turbo/Iterative Decoding suggests revisit of LDPC codes.
 - The decoders were feasible to implement 30 years later, reviving LDPC.
- 2020's LDPC codes find heavy use in modern designs.
 - 5G Wireless
 - Wi-Fi 5, 6, 7
 - High-speed Fiber
- Polar Codes (Arikan) 2009 (use another suboptimal "successive-decoding" method).
 - Even better for binary AWGN, but the BICM-ID does not work with PC's successive decoding and limits polar codes' applicability.



Section 8.3.3

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Some *H***-Related Definitions**

- 4 Cycle two rows have at least two 1's in same columns.
 - This is not good. Why?
 - Equality constraint \rightarrow to parity \rightarrow equality \rightarrow parity \rightarrow back again!
 - Biases accumulate quickly in constraint-based iterative decoding.

Regular Parity Matrix (sparse)

- All rows have t_r 1's.
- All columns have t_c 1's.
- So $(n-k) \cdot t_r = k \cdot t_c$.
- Otherwise, it is an irregular parity matrix.

Density-Limit bound:

- avoids all 4 cycles,
- ensures sparse *H* for finite *r*, &
- basically means *n* will be large.

$$r = 1 - \frac{t_c}{t_r}$$



$$n \leq \underbrace{\binom{(n-k)}{2}}_{\text{choose } 2 \text{ rows}} / \underbrace{\binom{t_c}{2}}_{\text{choose } 2 \text{ cols}} = \frac{(n-k) \cdot (n-k-1)}{t_c \cdot (t_c-1)}$$

Large n helps for "random coding" also ; so, how can a designer get such a code with implementable decoder? (typical n > 1000)



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Section 8.3.3

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Some LDPC Design Choices

Name	Quasi-Cyclic	Generic Irregular	Application Specific
Reg/irreg	regular	Slightly irregular	irregular
Uses	Wi-Fi	General	5G, DVB
positives	Matlab functions	379A class Matlab, no restrictions Good for M'ary	Puncturing Parallelism Special matlab
negatives	Not quite optimum	Not as well known/supported	Perhaps too specific

- There can be an SNR (equivalently *r*) dependence.
- Designers don't really want to design a new code for each channel.
- The code's amenability to puncturing/rate-variation is important.



Shaping Gain Offset

Review Lecture 6

- Turbo, LDPC, polar, ...
- DO NOT ADDRESS Shaping Gain
- See Section 8.5 for shaping codes
 - Can get up to 1.2 dB of the 1.53 dB
 - $\gamma_{s,offset}$ is shaping gain for particular constellation size (or \overline{b})

$\gamma \stackrel{\Delta}{=} rac{\left(d_{\min}^2(oldsymbol{x})/ar{\mathcal{E}}_{oldsymbol{x}} ight)}{\left(d_{\min}^2(oldsymbol{x})/ar{\mathcal{E}}_{oldsymbol{\check{x}}} ight)} =$	$\frac{\begin{pmatrix} \frac{d^2_{\min}(\boldsymbol{x})}{V^{2/N}(\Lambda)} \end{pmatrix}}{\begin{pmatrix} \frac{d^2_{\min}(\boldsymbol{\check{x}})}{V^{2/N}(\check{\Lambda})} \end{pmatrix}}_{\boldsymbol{\check{Y}_f}}_{fundamental}_{gain}}$	$\underbrace{ \begin{pmatrix} \frac{V^{2/N}(\Lambda)}{\bar{\mathcal{E}}_{\boldsymbol{\mathcal{X}}}} \end{pmatrix}}_{\begin{pmatrix} \frac{V^{2/N}(\check{\Lambda})}{\bar{\mathcal{E}}_{\check{\boldsymbol{\mathcal{X}}}}} \end{pmatrix}}_{\boldsymbol{\mathcal{Y}}_{S}} \\ \underbrace{ \boldsymbol{\mathcal{Y}}_{S}}_{\boldsymbol{\mathcal{S}} \boldsymbol{\mathcal{S}} \mathcal{S$	$\begin{cases} 0 < \gamma_s < 1.53 \end{cases}$
ar	This is what	t the LDPC/Tu	urbo works improve.
	Eb/N0 only	equals SNR v	when $r = 1/2$.

(t_c,t_r)	\overline{b}	γ_s offset	deviation from $\mathcal{C}_{ C =2}$
$(3,\!6)$.5	$.184 \mathrm{~dB}$	1.1 dB
(4,8)	.5	$.184 \mathrm{~dB}$	$1.6 \mathrm{dB}$
(5,10)	.5	$.184 \mathrm{~dB}$	$2.0 \mathrm{~dB}$
(3,5)	.4	$.051 \mathrm{~dB}$	$1.3 \mathrm{dB}$
(4,6)	1/3	$.033 \mathrm{~dB}$	1.4 dB

Regular codes' cannot get to capacity: Richardson/Urbanke, $\gamma_{s,offset}$ added here $\gamma_{s,offset} = \begin{cases} 0.1 \cdot \bar{b} \, \mathrm{dB} & 0 \leq \bar{b} \leq 0.33 \\ 0.27 \cdot \bar{b} - .057 \, \mathrm{dB} & 0.33 \leq \bar{b} \leq 0.4 \\ 1.33 \cdot \bar{b} - .48 \, \mathrm{dB} & 0.4 \leq \bar{b} \leq 0.5 \\ 0.2 \cdot \bar{b} + .084 \, \mathrm{dB} & 0.5 \leq \bar{b} \leq 1 \\ 1 \cdot \bar{b} - .72 \, \mathrm{dB} & 1 \leq \bar{b} \leq 2 \\ 0.2 \cdot \bar{b} + .85 \, \mathrm{dB} & 2 \leq \bar{b} \leq 3 \\ 0.17 \cdot \bar{b} + .83 \, \mathrm{dB} & 3 \leq \bar{b} \leq 4 \\ 1.53 \, \mathrm{dB} & \bar{b} \geq 4 \end{cases}$



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Section 8.3.3

Galois Field p reminder from L6:24-25

- $GF(p) = \{0, 1, \dots, p-1\}$
 - Addition is modulo *p*.
 - Multiplication is close, with division defined by inverse, and follows from any prime element $\alpha \in GF(p)$.

$$GF(5) = \{0 \ 1 \ \alpha \ \alpha^2 \ \alpha^3 \}$$

GF exists for any prime *p* **or product of such primes.**



×	1	2	3	4
1	1	2	3	4
2	2	4	1	3
3	3	1	4	2
4	4	3	2	1

α	α^2	α^3	α^4
2	4	3	1
3	4	2	1

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Appendix B.1

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Prelude to Quasi-Cyclic LDPC

- LDPC Design Goals:
 - Design avoids 4-cycles.
 - *H* should have rank n k.
 - *H* should have low density of 1's.
- Design should have good performance (including all neighbors at all distances),
 - but still have some structure to help encoder and especially decoder implementation.

 W is a Latin-Square Matrix. Each row/col contains each set element once. W is clearly nonsingular. 	$W = \begin{bmatrix} 0\\1\\2 \end{bmatrix}$	1 2 0	2 0 1
 J is a right-shift matrix (applied to row vector) that circularly shifts rows of matrix (on right) to the right, p×p. 	$J = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$	1 0 0	0 1 0
<i>H</i> is the dispersion of <i>W</i> using <i>J</i> over $GF(p)$ that • replaces each element $w_{i,j}$ by $(p-1) \times (p-1)$ matrix $J^{w_{i,j}}$.	$H = \begin{bmatrix} I \\ J \\ J^2 \end{bmatrix}$	J J ² I	J ² I]

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Not in text yet

Quasi-Cyclic LDPC

• Special Latin-Square $W(p \times p)$ matrix for any $\eta \in GF(p)$:

	$\eta - \alpha^0$	$\eta - lpha$		$\eta - \alpha^{p-2}$	η
	$\alpha \cdot \eta - \alpha^0$	$\alpha \cdot \eta - \alpha$		$\alpha \cdot \eta - \alpha^{p-2}$	$lpha\cdot\eta$
W =	÷	÷	÷	÷	:
	$\alpha^{p-2} \cdot \eta - \alpha^0$	$\alpha^{p-2}\cdot\eta-\alpha$		$\alpha^{p-2} \cdot \eta - \alpha^{p-2}$	$\alpha^{p-2}\cdot\eta$
	$-\alpha^0$	$-\alpha^1$		$-\alpha^{p-2}$	0

- W's dispersion (with (p 1)×(p 1) J) is the QC-LDPC matrix H and has no 4-cycles (Zhang et al.).
 Usually p = 2^m.
- These codes are regular (because of the *J* matrix and its shifts).
- Matlab ldpcQuasiCyclicMatrix.m command produces these:
 - Inputs are p-1 and W (which is a Latin-Square matrix with rules on how to create it).



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Not in text yet

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QC-LDPC codes and Sparse matrices



QC-LDPC encoder and decoder

- With the H matrix, create objects with ldpcEncoderConfig and ldpcDecoderConfig
 - Encode
 - Decode

>> wificonf=ldpcEncoderConfig(H) ParityCheckMatrix: [162 × 648 logical] Read-only properties: BlockLength: 648 NumInformationBits: 486 NumParityCheckBits: 162 CodeRate: 0.7500 wificonfdec=ldpcDecoderConfig(H,'norm-min-sum') ldpcDecoderConfig with properties: ParityCheckMatrix: [162 × 648 logical] Algorithm: 'norm-min-sum' **Read-only properties:** BlockLength: 648 NumInformationBits: 486 NumParityCheckBits: 162 CodeRate: 0.7500 NumRowsPerLayer: 27>> Y=ldpcEncode(X,wificonf);

>> X=prbs(7,486)';
>> Y=ldpcEncode(X,wificonf);
>> X1=ldpcDecode(1-2*Y,wificonfdec,6);
>> biterr(X,X1) = 0

>> error = [1 zeros(1,49) 1 zeros(1,49) 1 zeros(1,99) 1 zeros(1,45) 1 zeros(1,61)];
>> errorldpc=[error, error, zeros(1,32)];
>> X1=ldpcDecode(1-2*(Y+errorldpc'),wificonfdec,6);

>> biterr(X,X1) = 0

Warning: I could not get the 'bp' (Belief Propagation) option for IdpcDecode to work with noise UNLESS the errorIdpc/noise scales by <0.9; I think this relates to soft-info scaling internal to "bp" option

2nd decoder input can be 'bp', 'layered-bp', 'norm-min-sum', or 'offset-min-sum' and the corresponding algorithms are belief propagation decoding, layered belief propagation decoding, normalized min-sum decoding, and offset min-sum decoding respectively. https://www.mathworks.com/help/comm/ref/ldpcdecode.html

- You can begin to experiment now:
 - The decoder input is "LLR," so you could:
 - compute from a Gray mapped constellation,
 - run for different SNR,
 - compute error curves,
 - etc



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Not in text yet

Generic Irregular Codes

- Thanks go to E. Eleftheriou and S. Olcer of IBM (> 20 years so public domain ^(a)).
 - These use the shift-matrix dispersion concept and in easier way with $p \times p$ shift matrix J.
 - Their design checks for 4-cycles and linear-dependence \rightarrow irregular codes.
 - Their construction deletes any row that causes 4 cycle or linear dependence on previous rows.
 - The call the number of deleted rows m when the desired n k linearly independent rows is achieved.
- Starts with desired t_r and t_c

Eventually $n - k < t_c - p$

$$\tilde{t}_c \stackrel{\Delta}{=} (t_c - 1) \cdot m + t_c \cdot \left(\frac{n - m}{n}\right) \qquad r = 1 - \frac{\tilde{t}_c}{t_r}$$

$$H = \begin{bmatrix} I & I & \dots & I & I \\ I & J & J^2 & \dots & J^{t_r - 1} \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ I & J^{\tilde{t}_c - 1} & J^{2(\tilde{t}_c - 1)} & \dots & J^{(t_r - 1)(\tilde{t}_c - 1)} \end{bmatrix}$$

(n,k)	m	p	t_c	t_r	r	$\gamma_{s,offset} \ (8.71)$	Γ at 10^{-7}	$\gamma_{f,eff}$ at 10^{-7}
(276,209)	2	23	3	12	.7572	$0.69 \mathrm{dB}$	4.5 dB	5.1 dB
(529, 462)	2	23	3	23	.8733	$0.93~\mathrm{dB}$	$3.9~\mathrm{dB}$	$5.7~\mathrm{dB}$
(1369, 1260)	2	37	3	37	.9204	1.01 dB	$3.3~\mathrm{dB}$	$6.3~\mathrm{dB}$
(2209, 2024)	3	47	4	47	.9163	1.01 dB	$2.8~\mathrm{dB}$	$6.8~\mathrm{dB}$
(4489, 4158)	4	67	5	67	.9263	1.03 dB	$2.5~\mathrm{dB}$	7.1 dB
(7921, 7392)	5	89	6	89	.9332	1.04 dB	$2.3~\mathrm{dB}$	$7.3~\mathrm{dB}$

Table 8.21: Generic LDPC code parameters.



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Sec 8.3.3.3

Generic Software (customized to 379A)

To get H (not yet in sparse format)

function [H_no_dep H] = get_h_matrix(p,tr,tc,first_1); Generate LDPC H Matrix Uses Generic-LDPC Method As Per Cioffi's Class Notes Example: to Generate (529,462) code, p=23, rw=23, cw=3, first_1=2 H = get_h_matrix(23,23,3,2);

Definition of input variables

- p : Prime number of the size of base matrix of size p-by-p
- tr : Row weight = # of base matrices (or 1's) /row, equivalent to K

tc : Col weight = # of base matrices (or 1's) per column,eq to J first_1: Set to 2 in generic LDPC code, so right shift by first_1-1

Definition of output variables

H_no_dep : the parity check matrix with no dependent rows

H : without removing the dependent rows

EE379A, Chien-Hsin Lee, first version 06/2006, edits by J. Cioffi since

```
>> H = get_h_matrix(23,23,3,2);
>> size(H) = 67 529
>> 529-67 = 462
>> H=nonsinglastnk(H);
>> generic=ldpcEncoderConfig(logical(sparse(H)))
ParityCheckMatrix: [67 × 529 logical]
BlockLength: 529
NumInformationBits: 462
NumParityCheckBits: 67
CodeRate: 0.8733
```

>> X=prbs(7,462);

- >> Y=ldpcEncode(X',generic);
- >> genericdec=ldpcDecoderConfig(generic,"norm-min-sum");
- >> errorgeneric=[error, 1 zeros(1, 99), 1 1 zeros(1,98) zeros(1,21)];
- >> size(errorgeneric) % = 1 529
- >> X1=ldpcDecode(1-2*(Y+1*errorgeneric'),genericdec,6);
- >> biterr(X',X1) % = 0



Stanford University

PS5.3 (8.14)

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Other Irregular

- Digital Video Broadcast standard has:
 - r =1/4, 1/3, 2/5, 1/2, 3/5, 2/3, 3/4, 4/5, 5/6, 8/9, or 9/10
 - *n* = 64,800
 - Can then use ldpcencode.m and ldpcdecode.m .
- 5G standard (for 5G's live data, not %G's control channel):
 - has good puncturing, parallelism, and gain (see slides 27,28)
 - Is specific to this application, but may be good elsewhere also.
 - Matlab commands are
 - >> nrLDPCEncode.m >> nrLDPCDecode.m



>> Hdvb=dvbs2ldpc(r)

5G Code

- Using same "lifting" (Latin Squares) except with allzeros matrices also allowed in some positions (so J → {J, 0} = Z).
 - Many forms of the Z matrices to be lifted that use two "base matrices."
- Former 379 student Rick Wesel (now UCLA Prof) contributed concepts that allow:
 - Scalable decoder complexity with rate choice over wide range from 1/5 to 1/3
 - See reference [7] in Ericsson article below.
- See tutorial articles by
 - Qualcomm: Tom Richardson and Shrinivas Kudekar, "Design of Low-Density Parity Check Codes for 5G New Radio." <u>IEEE Communications Magazine</u> (Volume: 56, <u>Issue: 3</u>, March 2018), **pp.** 28 - 34, **DOI:** https://ieeexplore.ieee.org/document/8316763.
 - Ericsson: Dennis Hui et al, "Channel Coding in 5G New Radio," <u>IEEE Vehicular Technology Magazine</u> (Volume: 13, <u>Issue: 4</u>, December 2018), 60 - 69, DOI: <u>10.1109/MVT.2018.2867640</u>.
- More parity bits sent upon CRC failure (see L11).
 - Complexity scales with N (rate increase)
 - Unlike puncturing with turbo codes





FIGURE 2 The structure of NR LDPC base matrix 1. Each square corresponds to one element in the base matrix or a $Z \times Z$ subblock in the PCM.



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More 5G codes (Ericsson paper)



TABLE 1 NR LDPC base matrix parameters.

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Parameter	Base Matrix 1	Base Matrix 2
Minimum design code rate	1/3	1/5
Base matrix size	46 × 68	42 × 52
Number of system- atic columns	22	10
Maximum informa- tion block size <i>K</i>	8,448 (= 22 × 384)	3,840 (= 10 × 384)
Number of nonzero elements	316	197

FIGURE 1 The NR LDPC coding chain.

5G mandates base code use by rate and K



FIGURE 3 The usage of the two base matrices specified for the NR data channel. For K larger than the maximum information block size, code block segmentation is applied.



FIGURE 4 The performance of NR LDPC codes at code rate 1/2 for QPSK modulation.



Polar Codes

Section 8.3.4

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Polar Codes Brief Commentary

- Polar Codes Positives (Arikan); PC are:
 - not random,
 - based on essentially finite-field Fourier Transform size n,
 - have simpler suboptimal decoders (successive decoders),
 - smaller gap for finite n,
 - used for binary (BPSK) control channel in 5G, &
 - lower delay.

- Polar Codes Negatives:
 - Code design strong depends on SNR, instead of puncturing.
 - The successive decoder is not really compatible with M'ary QAM.
 - PC don't provide that much more gain.



Not (yet) heavily used

So not in 379's

There are much bigger impacts to performance that arise from a). Handling ISI/filtering – A and B b). Optimizing transmit spectra, B c). Allocation of dimensions/energy to multiple users sharing channel (B)

GRAND Decoders (L12) get same or better gain for simple block codes used as product codes, with yet lower decoder computation.



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Not in text yet.

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End Lecture 10

backup

$$p(v_i) = \begin{cases} Pr\{v_i = 0\} = \frac{(1 - a_1) \cdot (1 - a_2) \cdot (1 - a_3)}{a_1 \cdot a_2 \cdot a_3 + (1 - a_1) \cdot (1 - a_2) \cdot (1 - a_3)} \\ Pr\{v_i = 1\} = \frac{a_1 \cdot a_2 \cdot a_3}{a_1 \cdot a_2 \cdot a_3 + (1 - a_1) \cdot (1 - a_2) \cdot (1 - a_3)} \end{cases}$$

sd sd

