

LDPC Codes: Achieving the Capacity of the Binary Erasure Channel

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Nov. 30, 2009

Outline

- 1 Coding and the BEC
- 2 LDPC Codes
- 3 Density Evolution
- 4 Irregular LDPC Codes
- 5 Conclusion

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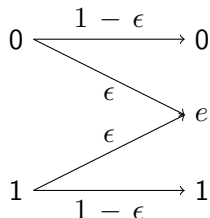
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Error Control Codes

- Consider a data transmission system whereby binary data is segmented into *messages* \mathbf{u} of length k bits.
- Each message is mapped to a unique *codeword* \mathbf{c} of length n bits, where $n > k$.
- The ratio $R = k/n$ is called the *code rate*.
- Simple examples:
 - Repetition code: $k = 1$; Repeat bit n times; $R = 1/n$.
 - Single parity-check code: Codeword is the message and an additional “parity bit”; $n = k + 1$; $R = k/(k + 1)$.

The Binary Erasure Channel

- The BEC has two inputs (data 0 and data 1) and three outputs (data 0, data 1, and *erasure* e).
- A bit is erased with probability ϵ .
- A bit is correctly received with probability $1 - \epsilon$.



- Example applications:
 - Buffer overflows in network routers.
 - Fading in wireless channels.

Capacity of the BEC

- According to information theory, it is possible to reliably communicate over the BEC by using a rate $R = 1 - \epsilon$ code.
- Can be easily achieved if the *transmitter* knows the location of the erasures.
- Example: Transmit $\mathbf{u} = [1 \ 0 \ 1 \ 1]$ with a rate $R = 4/6$ code:

$$\mathbf{c} = \left[\underbrace{1}_{c_1=u_1} \ \underbrace{e}_{c_2=X} \ \underbrace{0}_{c_3=u_2} \ \underbrace{e}_{c_4=X} \ \underbrace{1}_{c_5=u_3} \ \underbrace{1}_{c_6=u_4} \right]$$

where X can be anything (does not matter, since erased).

- This scheme is not practical, since normally the transmitter won't know where the erasures are located, and therefore doesn't know where to place the message bits.
- Finding practical codes which require only the *receiver* to know the location of the erasures is a challenging problem.

Single Parity-Check Codes

- Consider the following rate $R = 5/6$ parity-check code:

$$\mathbf{c} = [\underbrace{1 \ 0 \ 1 \ 0 \ 1}_{\mathbf{u}} \ \underbrace{1}_{\text{parity bit}}]$$

- One erasure in *any* position may be corrected:

$$\mathbf{c} = [1 \ 0 \ e \ 0 \ 1 \ 1]$$

- Problem with using SPC's is that it can only correct a single erasure.

Product Codes: Encoding

- Place data into a k by k rectangular array.
 - Encode each row with a SPC.
 - Encode each column with a SPC.
 - Result is a rate $R = k^2/(k+1)^2$ code.
- Example $k = 2$.

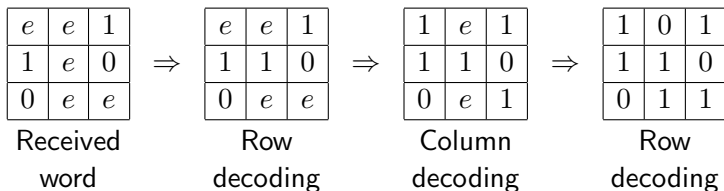
$c_1 = u_1$	$c_2 = u_2$	$c_3 = c_1 \oplus c_2$
$c_4 = u_3$	$c_5 = u_4$	$c_6 = c_4 \oplus c_5$
$c_7 = c_1 \oplus c_4$	$c_8 = c_2 \oplus c_5$	$c_9 = c_3 \oplus c_6$

=

1	0	1
1	1	0
0	1	1

Product Codes: Decoding

- Decoding may be performed by iteratively decoding the SPC on each row and column.



- Does not achieve capacity. Try decoding:

e	e	1
e	e	0
0	1	1

Linear Codes

$c_1 = u_1$	$c_2 = u_2$	$c_3 = c_1 \oplus c_2$
$c_4 = u_3$	$c_5 = u_4$	$c_6 = c_4 \oplus c_5$
$c_7 = c_1 \oplus c_4$	$c_8 = c_2 \oplus c_5$	$c_9 = c_3 \oplus c_6$

- The example product code is characterized by the set of five linearly-independent equations:

$$c_3 = c_1 \oplus c_2 \Rightarrow c_1 \oplus c_2 \oplus c_3 = 0$$

$$c_6 = c_4 \oplus c_5 \Rightarrow c_4 \oplus c_5 \oplus c_6 = 0$$

$$c_7 = c_1 \oplus c_4 \Rightarrow c_1 \oplus c_4 \oplus c_7 = 0$$

$$c_8 = c_2 \oplus c_5 \Rightarrow c_2 \oplus c_5 \oplus c_8 = 0$$

$$c_9 = c_3 \oplus c_6 \Rightarrow c_3 \oplus c_6 \oplus c_9 = 0$$

- In general, it takes $(n - k)$ linearly-independent equations to specify a *linear code*.

Parity-check Matrices

- The system of equations may be expressed in matrix form as:

$$\mathbf{c}H^T = \mathbf{0}$$

where H is a *parity-check* matrix.

- Example:

$$\begin{array}{l}
 c_1 \oplus c_2 \oplus c_3 = 0 \\
 c_4 \oplus c_5 \oplus c_6 = 0 \\
 c_1 \oplus c_4 \oplus c_7 = 0 \\
 c_2 \oplus c_4 \oplus c_8 = 0 \\
 c_3 \oplus c_6 \oplus c_9 = 0
 \end{array}
 \Leftrightarrow
 H =
 \begin{bmatrix}
 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\
 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\
 0 & 1 & 0 & 1 & 0 & 0 & 0 & 1 & 0 \\
 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1
 \end{bmatrix}$$

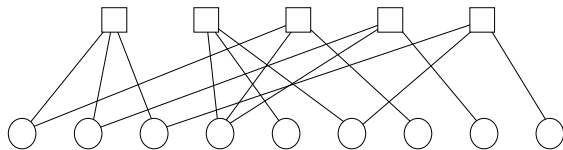
Parity-check matrix

Tanner Graphs

- The parity-check matrix may be represented by a *Tanner* graph.
- Bipartite graph:
 - Check nodes: Represent the $n - k$ parity-check equations.
 - Variable nodes: Represent the n code bits.
- If $H_{i,j} = 1$, then i^{th} check node is connected to j^{th} variable node.
- Example: For the parity-check matrix:

$$H = \begin{bmatrix} 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$

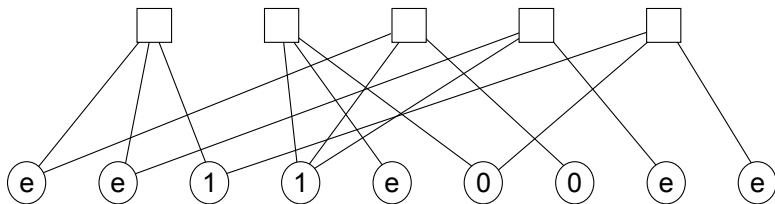
The Tanner Graph is:



Decoding on the Tanner Graph

Decoding can be performed on the Tanner graph.

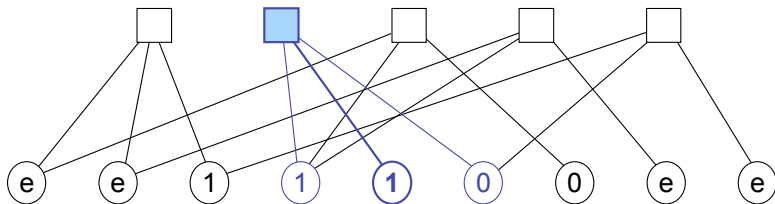
- Load the variable nodes with the observed code bits.
- Each check node j sends a *message* to each of its connected variable nodes i .
 - The message is the modulo two sum of the bits associated with the connected variable nodes *other* than i (if none are erased).
 - If a check node touches a *single* erasure, then it will become corrected.
- Iterate until all erasures corrected or no more corrections possible.



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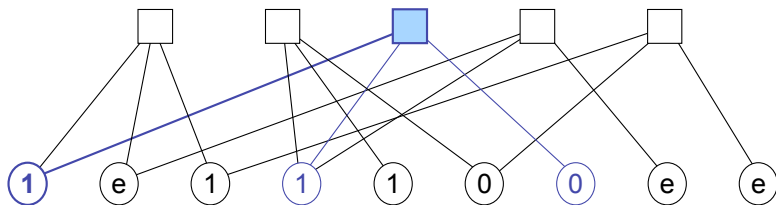
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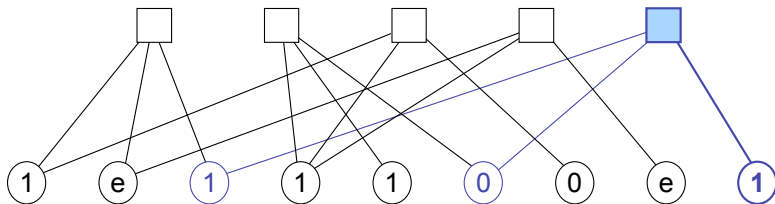
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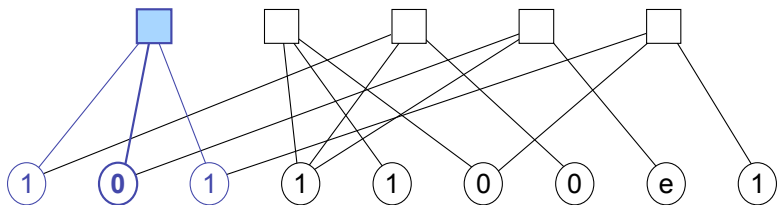
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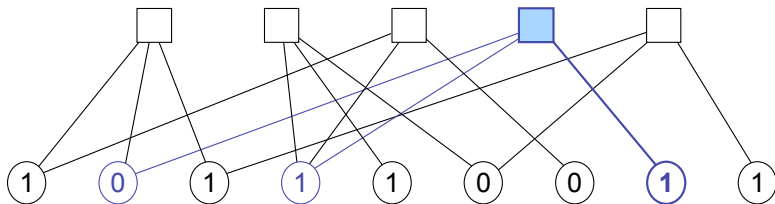
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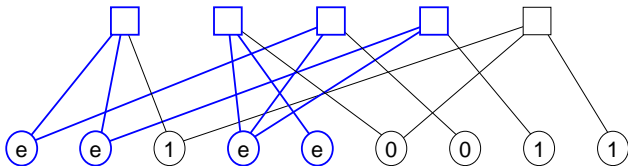
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Stopping sets

- A *stopping set* \mathcal{V} is a set of erased variable nodes that cannot be corrected, regardless of the state of the other variable nodes.



- Let \mathcal{G} be the neighbors of \mathcal{V} .
- Every check node in \mathcal{G} touches at least two variable nodes in \mathcal{V} .
- The *minimum* stopping set \mathcal{V}_{min} is the stopping set containing the fewest variable nodes.
- Let $d_{min} = |\mathcal{V}_{min}|$ be the size of the minimum stopping set.
 - There exists at least one pattern of d_{min} erasures that cannot be corrected.
 - The erasure correcting capability of the code is $d_{min} - 1$, which is the maximum number of erasures that can always be corrected.

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LDPC Codes

- Observations:
 - The decoder's complexity depends on the degree of the check nodes.
 - The degree of a check node is equal to the Hamming weight of the corresponding row of the parity-check matrix.
 - To achieve capacity, a long code is needed.
 - It is desirable to have a code that is long, yet with small row weight.
- Low-density parity-check codes:
 - An LDPC code is characterized by a *sparse* parity-check matrix.
 - The row/column weights are independent of length.
 - Decoder complexity grows only linear with block length.
- Historical note:
 - LDPC codes were the subject of Robert Gallager's 1960 dissertation.
 - Were forgotten because the decoder could not be implemented.
 - Were "rediscovered" in the mid-1990's after turbo codes were developed.

Example LDPC Code

- A code from MacKay and Neal (1996):

$$\mathbf{H} = \left[\begin{array}{cccc|ccc|ccc}
 1 & & & & & 1 & & & & 1 & & & \\
 1 & & & & 1 & & & & & & 1 & & \\
 & & 1 & & 1 & & & & & & & 1 & \\
 & & & 1 & & & 1 & & & & & & 1 \\
 & & & & 1 & & & 1 & & & & & 1 \\
 & & 1 & & & & & & & 1 & & & \\
 1 & & & 1 & & & 1 & & & & 1 & & \\
 & & 1 & & & & & & & & & 1 & \\
 & & & 1 & 1 & & & & & 1 & & & \\
 & & & & & 1 & & & & & 1 & & \\
 & & & & & & 1 & & & & & 1 & \\
 & & & & & & & & & 1 & & & 1
 \end{array} \right]$$

- The code is *regular* because:
 - The rows have constant weight (check-nodes constant degree).
 - The columns have constant weight (variable-nodes constant degree).
- This is called a (3,4) regular code because the variable nodes have degree 3 and the check nodes have degree 4.

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Density Evolution

- For a (d_v, d_c) regular code, the probability that a variable-node remains erased after the ℓ^{th} iteration is

$$\epsilon_\ell = \epsilon_0 \left(1 - (1 - \epsilon_{\ell-1})^{d_c-1} \right)^{d_v-1} \quad (1)$$

where d_v is the variable-node degree, d_c is the check-node degree, and the initial condition is $\epsilon_0 = \epsilon$.

- The above result assumes independent messages, which is achieved when the girth of the Tanner graph is sufficiently large.
- If $\epsilon_\ell \rightarrow 0$ as $\ell \rightarrow \infty$ for a particular channel erasure probability ϵ , then a code drawn from the ensemble of all such (d_v, d_c) regular LDPC codes will be able to correctly decode.
- The *threshold* ϵ^* is the maximum ϵ for which $\epsilon_\ell \rightarrow 0$ as $\ell \rightarrow \infty$.
- For the $(3, 6)$ regular code, the threshold is $\epsilon^* = 0.4294$

Proof of (1), Part I/II

- Decoding involves the exchange of messages between variable nodes and check nodes.
 - Let p_{\uparrow} denote the probability of an erased message going *up* from the variable nodes to the check nodes.
 - Let p_{\downarrow} denote the probability of an erased message going *down* from the check nodes to the variable nodes.
- Consider the degree d_c check node.
 - An outgoing message sent over a particular edge is a function of the incoming messages arriving over the other $d_c - 1$ edges.
 - For the outgoing message to be correct, all $d_c - 1$ incoming messages must be correct.
 - The outgoing message will be an erasure if any of the $d_c - 1$ incoming messages is an erasure.
 - The probability of the check node sending an erasure is:

$$p_{\downarrow} = 1 - (1 - p_{\uparrow})^{d_c - 1} \quad (2)$$

Proof of (1), Part II/II

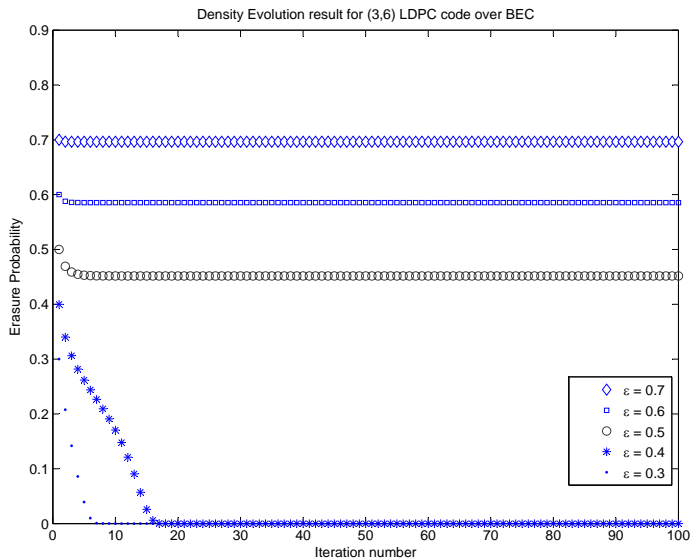
- Consider the degree d_v check node.
 - An outgoing message sent over a particular edge is a function of the incoming messages arriving over the other $d_v - 1$ edges.
 - An outgoing message will be an erasure if the variable node was initially erased *and* all of the arriving messages are erasures.
 - The probability of the variable node sending an erasure is:

$$p_{\uparrow} = \epsilon_0 p_{\downarrow}^{d_v - 1} \quad (3)$$

- Letting ϵ_{ℓ} equal the value of p_{\uparrow} after the ℓ^{th} iteration, and substituting (2) into (3) yields the recursion given by (1):

$$\epsilon_{\ell} = \epsilon_0 \left(1 - (1 - \epsilon_{\ell-1})^{d_c - 1} \right)^{d_v - 1}$$

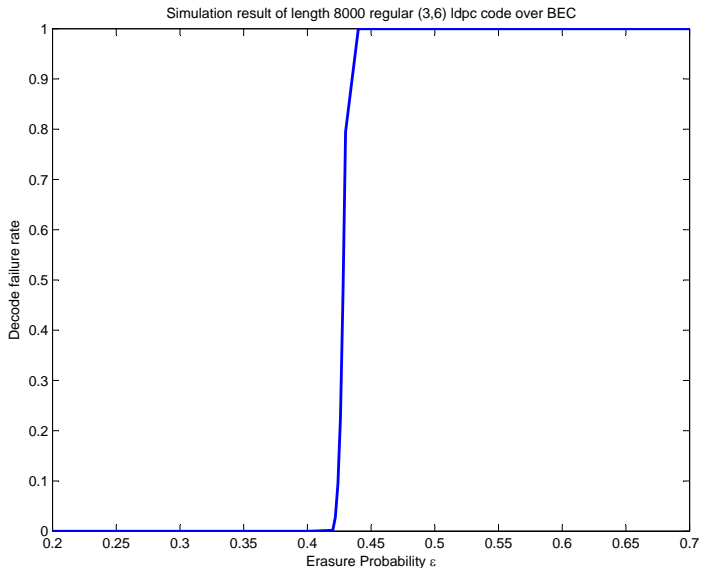
DE Example



Code Realization

- Density evolution only describes the asymptotic performance of the *ensemble* of LDPC codes.
- Implementation requires that an H matrix be generated by drawing from the ensemble of all (d_v, d_c) LDPC codes.
- Goals of good H design:
 - High girth.
 - Full rank.
 - Large minimum stopping set.
- If the girth is too low, the short cycles invalidate the iterative decoder.
- High girth achieved through girth conditioning algorithms such as progressive edge growth (PEG).
- If H is not full rank, then the rate will be reduced according to the number of dependent equations.
- Small stopping sets give rise to an *error floor*.
- A database of good regular LDPC codes can be found on MacKay's website.

Performance of an Actual Code



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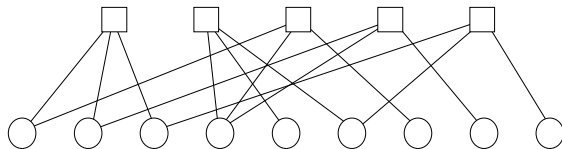
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Irregular LDPC Codes

- Although regular LDPC codes perform well, they are not capable of achieving capacity.
- Properly designed irregular LDPC codes are capable of achieving capacity.
 - The degree distribution of the variable nodes is not constant.
 - The check-node degrees are often still constant (or close to it).
 - Here “designing” means picking the proper degree distribution.

Degree Distribution

- Edge-perspective degree distributions:
 - ρ_i is the fraction of *edges* touching degree i check nodes.
 - λ_i is the fraction of *edges* touching degree i variable nodes.
- For example, consider the Tanner graph:



- 15 edges.
- All are connected to degree-3 check nodes, so $\rho_3 = 15/15 = 1$.
- Four are connected to degree-1 variable nodes, so $\lambda_1 = 4/15$.
- Eight are connected to degree-2 variable nodes, so $\lambda_2 = 8/15$.
- Three are connected to the degree-3 variable node, so $\lambda_3 = 3/15$.

DE for Irregular LDPC

- The degree distributions are described in polynomial form:
 - $\rho(x) = \sum_i \rho_i x^{i-1}$ for check nodes.
 - $\lambda(x) = \sum_i \lambda_i x^{i-1}$ for variable nodes.
- For an irregular code, the probability that a variable-node remains erased after the ℓ^{th} iteration is

$$\epsilon_\ell = \epsilon_0 \lambda(1 - \rho(1 - \epsilon_{\ell-1}))$$

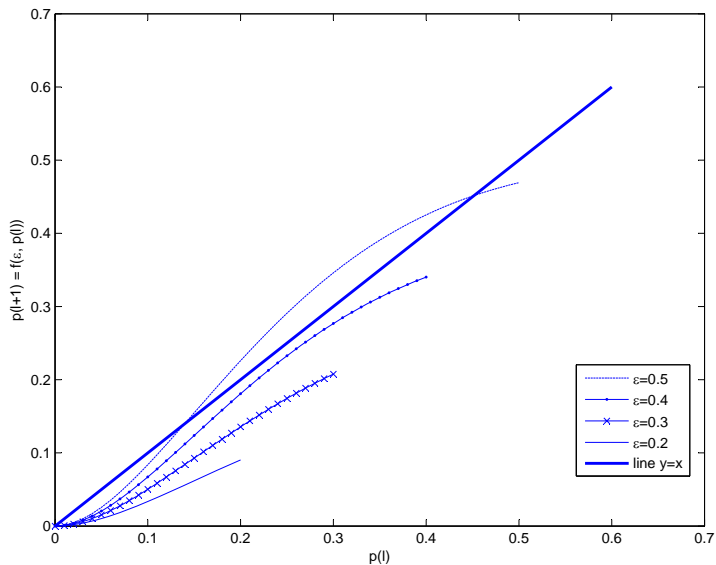
The proof follows from the Theorem on Total Probability.

- Convergence:
 - Error-free decoding requires that the erasure probability goes down from one iteration to the next.
 - Define the related function:

$$f(\epsilon, x) = \epsilon \lambda(1 - \rho(1 - x))$$

- Error-free decoding is possible iff $f(\epsilon, x) \leq x$ for all $0 \leq x \leq \epsilon$.

Convergence



Optimization

- The threshold is

$$\epsilon^* = \sup\{\epsilon : f(\epsilon, x) < x, \forall x, 0 < x \leq \epsilon\}$$

- Solving $f(\epsilon, x) = x$ for ϵ

$$\begin{aligned} x &= f(\epsilon, x) \\ &= \epsilon\lambda(1 - \rho(1 - x)) \\ \epsilon &= \frac{x}{\lambda(1 - \rho(1 - x))} \end{aligned}$$

which is a function of x , and henceforth expressed as $\epsilon(x)$.

- This allows the threshold to be rewritten as:

$$\epsilon^* = \min\{\epsilon(x) : \epsilon(x) \geq x\}$$

Optimization with Linear Programming

- Our goal is to find the degree distribution which yields maximum threshold

$$\max_{\varepsilon^*} \{ \varepsilon^* = \min(\varepsilon(x) : \varepsilon(x) \geq x) \};$$

- Several Constraints

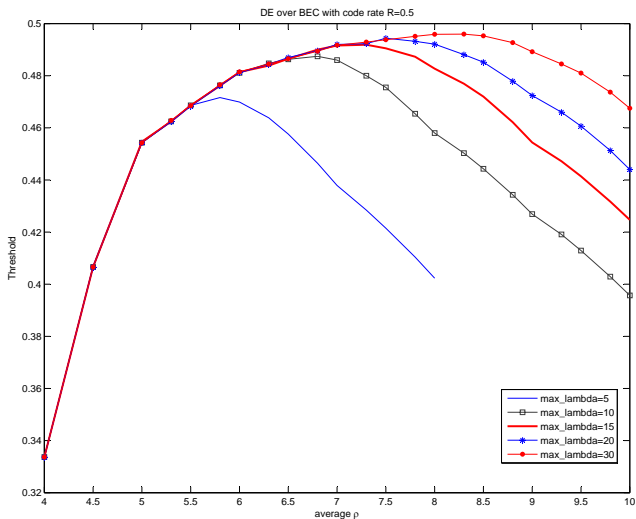
$$\frac{\int_0^1 \rho(x) dx}{\int_0^1 \lambda(x) dx} = 1 - R$$

$$\sum_{i \geq 2} \lambda_i = 1; \quad \sum_{i \geq 2} \rho_i = 1;$$

$$x \in [0, 1]$$

- Which can be modeled as a optimization problem using linear programming
 - Can use Matlab's Optimization Toolbox.

Optimization Results ($\epsilon^* = 0.49596$)



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Conclusion

- Conclusions:
 - Irregular LDPC codes can achieve the capacity of the BEC channel.
 - Density evolution predicts asymptotic performance.
 - Key to design is picking the degree distributions.
- Related Issues:
 - Predicting performance of finite-length codes (and designing them).
 - Dealing with unknown ϵ (rateless coding).
 - Dealing with other channels (AWGN, etc.).
- A plug:
 - EE 567: Coding Theory.
 - T/H 5:00-6:15 PM on Evansdale Campus.
 - Will cover linear codes in general and LDPC codes in particular.
 - All you need is graduate-level mathematical maturity and a sense of inquisitiveness.

Thank You.