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# Short Variable-Length Codes with Shared Incremental Redundancy: A New Architecture for High Throughput

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2019 Oberpfaffenhofen Workshop on High Throughput Coding

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UCLA Electrical and Computer Engineering Department

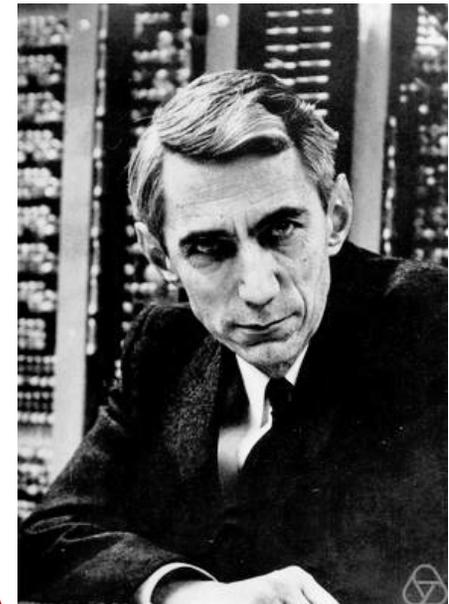
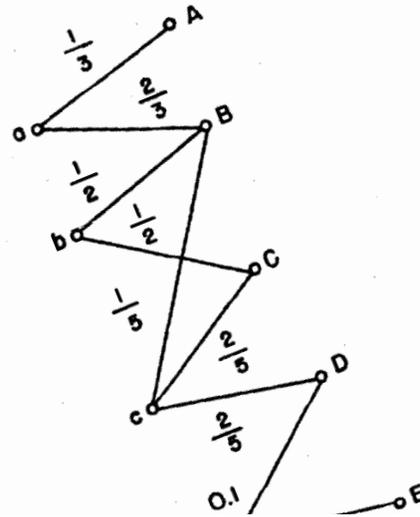
# Shannon: Feedback does not increase capacity

## THE ZERO ERROR CAPACITY OF A NOISY CHANNEL

Claude E. Shannon  
Bell Telephone Laboratories, Murray Hill, New Jersey  
Massachusetts Institute of Technology, Cambridge, Mass.

### Abstract

The zero error capacity  $C_0$  of a noisy channel is defined as the least upper bound of rates at which it is possible to transmit information with zero probability of error. Various properties of  $C_0$  are studied; upper and lower bounds and methods of evaluation of  $C_0$  are given. Various "products" of two given channels,  $C_0$  of a feedback link is considered. It is shown that the ordinary capacity of a channel with feedback is equal to the ordinary capacity of the channel without feedback. The solution



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## Shannon: Feedback does not increase capacity

Mutual information between message  $W$   
and receiver's knowledge

Before we send  $X_j = f(W, Y_1 \cdots Y_{j-1})$ :

$$I(W; Y_1^{j-1})$$

After we send  $X_j = f(W, Y_1 \cdots Y_{j-1})$ :

$$I(W; Y_1^{j-1}, Y_j)$$



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$$I(W; Y_1^{j-1}, Y_j) - I(W; Y_1^{j-1}) \leq C$$

# An Operational Interpretation and a surprise.

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## THEOREM (SHANNON)

Any rate that can be achieved with feedback can be achieved without feedback.

## COROLLARY (Further constraining the feedback cannot help.)

Any rate that can be achieved with ACK/NACK feedback and a Variable-Length Code can be achieved without feedback.

## SURPRISE

Any rate that can be achieved with ACK/NACK feedback and a Variable-Length Code can be **closely approached** without feedback **using the same Variable-Length Code**.



# ITA 2019 Presentation Final Slide

$$\tau^* = \min_n \{n \geq 0 : i(X^n(\theta), Y^n) > \gamma\}$$



Abraham Wald

$$E\tau^* < \frac{\gamma}{C} + 1$$



Theorem 3: Fix a scalar  $\gamma > 0$ , a channel  $\{P_{Y_i|X_i, Y^{i-1}}\}_{i=1}^{\infty}$  and an arbitrary process  $X = (X_1, X_2, \dots, X_n, \dots)$  taking values in  $\mathcal{A}$ . Define a probability space with finite-dimensional distributions given by (23).  $X$  and  $\tilde{X}$  are independent copies of the same process and  $Y$  is the output of the channel when  $X$  is its input. For the joint distribution (23)  $A^n \times B^n \rightarrow \mathbb{R}$  define a sequence of information density functions

$$u(a^n; b^n) = \log \frac{dP_{Y^n|X^n}(b^n|a^n)}{dP_{Y^n}(b^n)} \quad (24)$$

and a pair of hitting times

$$\begin{aligned} \tau &= \inf\{n \geq 0 : u(X^n; Y^n) \geq \gamma\} \\ \tilde{\tau} &= \inf\{n \geq 0 : u(\tilde{X}^n; Y^n) \geq \gamma\}. \end{aligned} \quad (25)$$

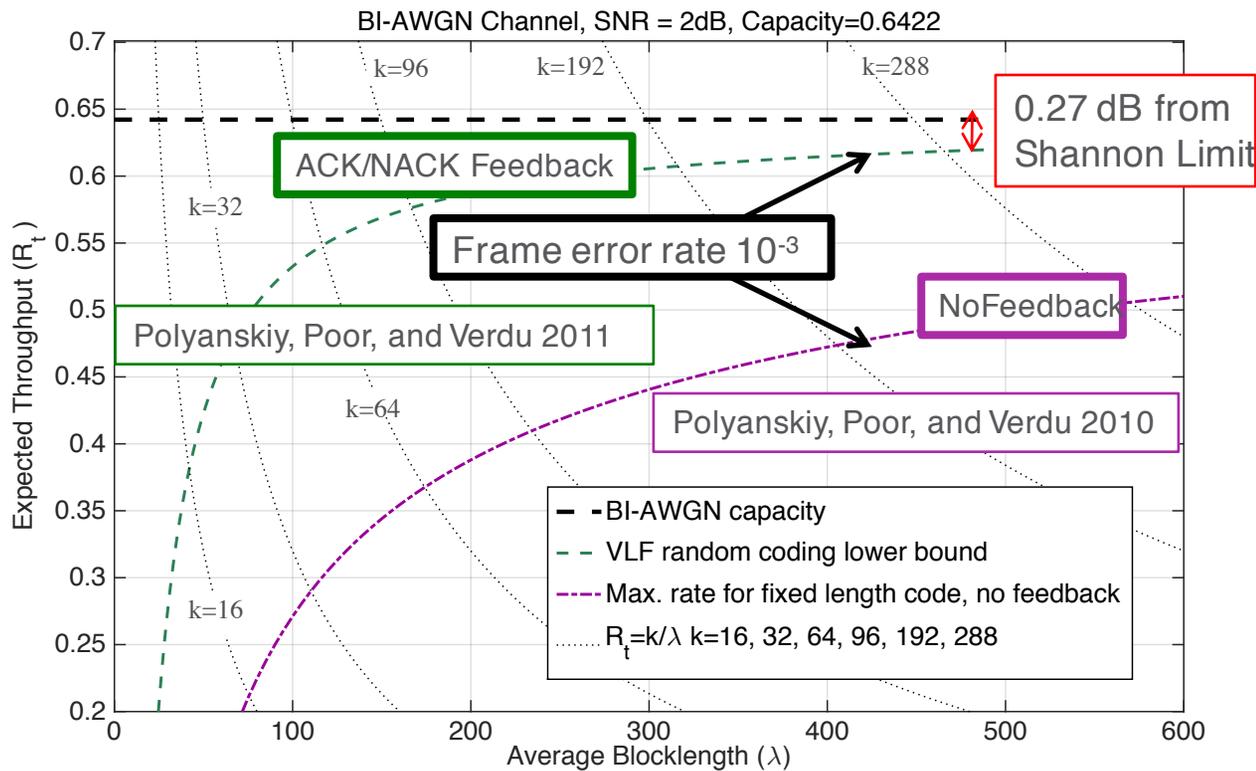
Then for any  $M$  there exists an  $(\ell, M, \epsilon)$  VLF code with

$$\begin{aligned} \ell &\leq E[\tau] \\ \epsilon &\leq (M-1)\mathbb{P}[\tilde{\tau} \leq \tau]. \end{aligned}$$

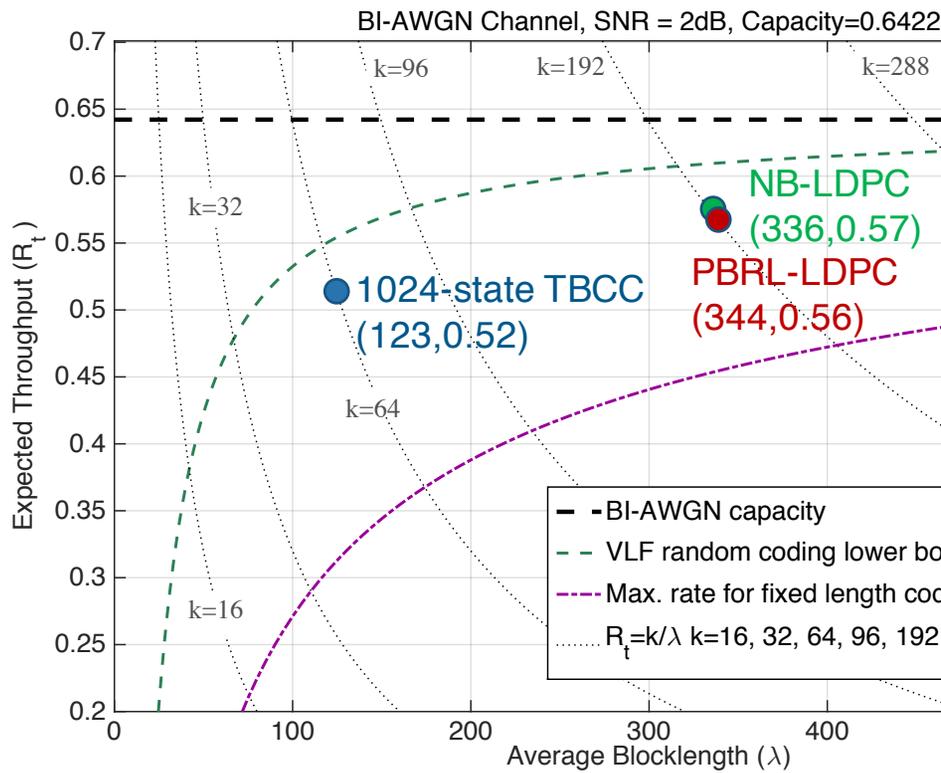
Furthermore, for any  $M$  there exists a deterministic VLF code with  $\epsilon$  satisfying (28) and

$$\ell \leq \text{esssup } E[\tau|X].$$

# PPV Achievable Rates for BI-AWGN Channel



# Real codes, 5 Transmissions, Constant



IEEE TRANSACTIONS ON COMMUNICATIONS, VOL. 63, NO. 7, JULY 2015

Variable-Length Convolutional Coding for Short Blocklengths With Decision Feedback

IEEE TRANSACTIONS ON COMMUNICATIONS, VOL. 64, NO. 6, JUNE 2016

Optimizing Transmission Lengths for Limited Feedback With Nonbinary LDPC Examples

2017 IEEE International Symposium on Information Theory (ISIT)

Approaching Capacity Using Incremental Redundancy without Feedback

IEEE TRANSACTIONS ON COMMUNICATIONS, VOL. 63, NO. 5, MAY 2015

Protograph-Based Raptor-Like LDPC Codes

1522  
This article has been accepted for publication in a future issue of this journal, but has not been fully edited. Content may change prior to final publication. Citation information: DOI 10.1109/TIT.2019.2895322, IEEE Transactions on Information Theory

Quasi-Cyclic Protograph-Based Raptor-Like LDPC Codes for Short Block-Lengths

V. S. Ranganathan, Student Member, IEEE, Dariush Divsalar, Life Fellow, IEEE, and Richard D. Wesel, Senior Member, IEEE

of a certain size over the channel. If the receiver is unable to decode the received noisy codeword, the transmitter retransmits additional codeword symbols to the receiver to increase the total number of received symbols. The process continues until the received codeword can be decoded. The process continues until the received codeword can be decoded.

# An Operational Interpretation and a surprise.

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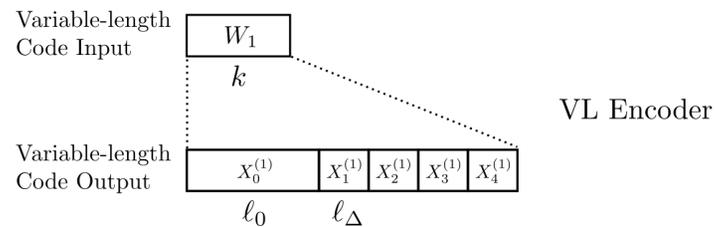
## SURPRISE

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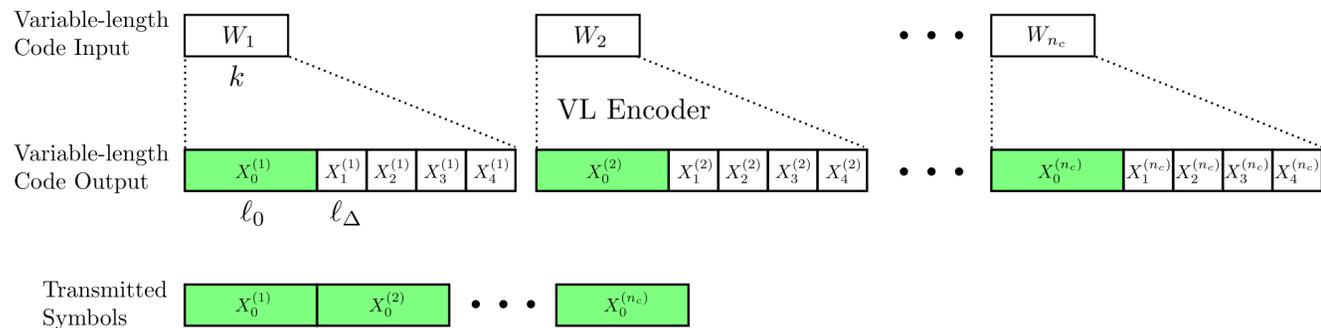
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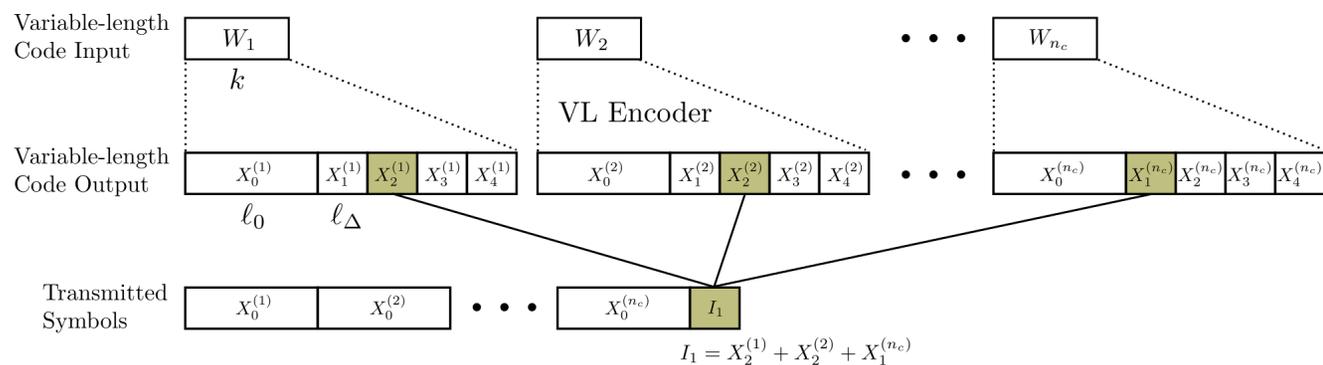
# Inter-frame Coding [Zeineddine & Mansour]



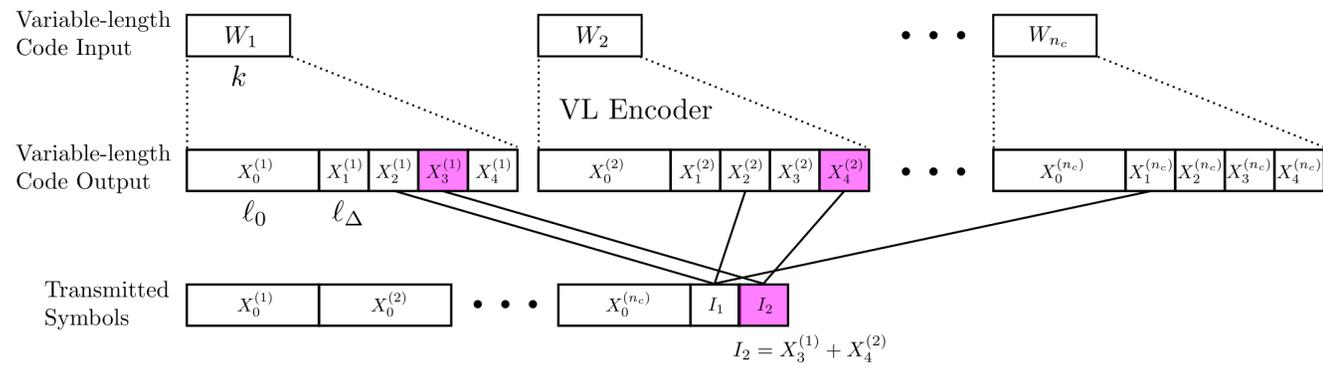
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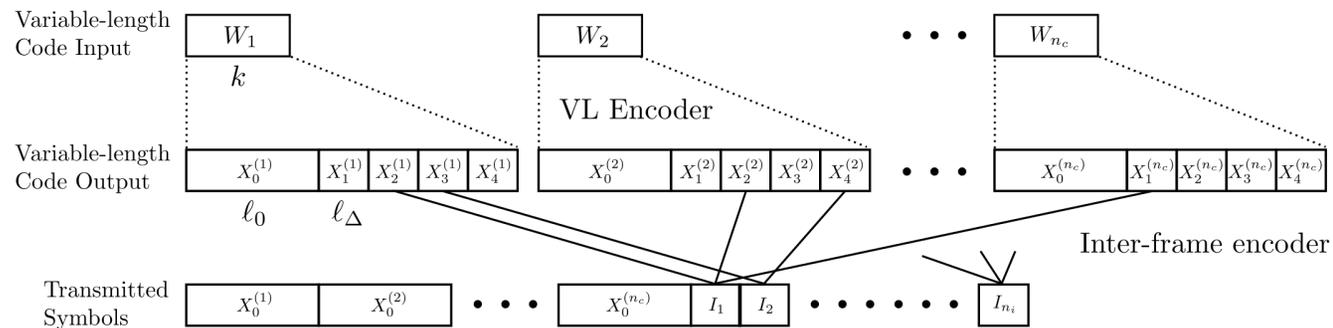
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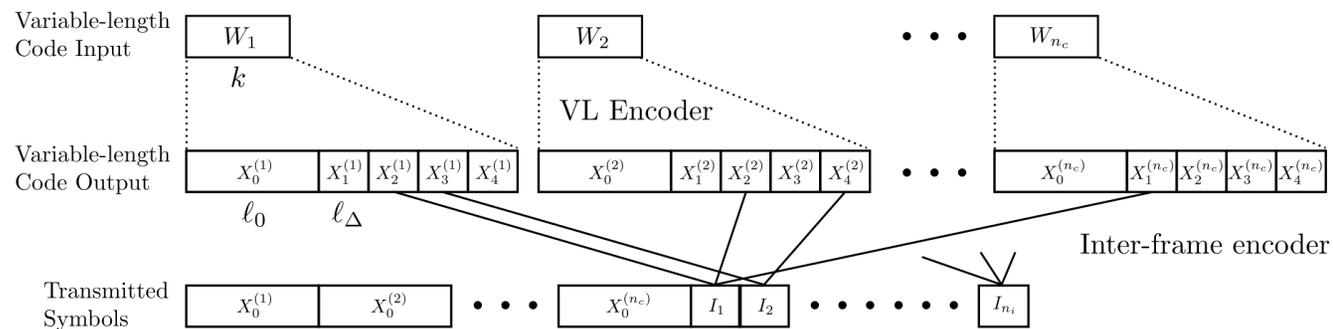
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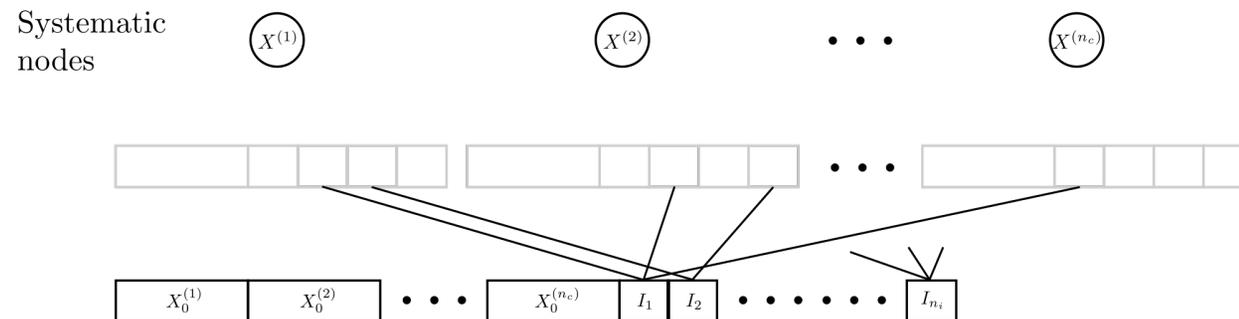
# Inter-frame Coding [Zeineddine & Mansour]



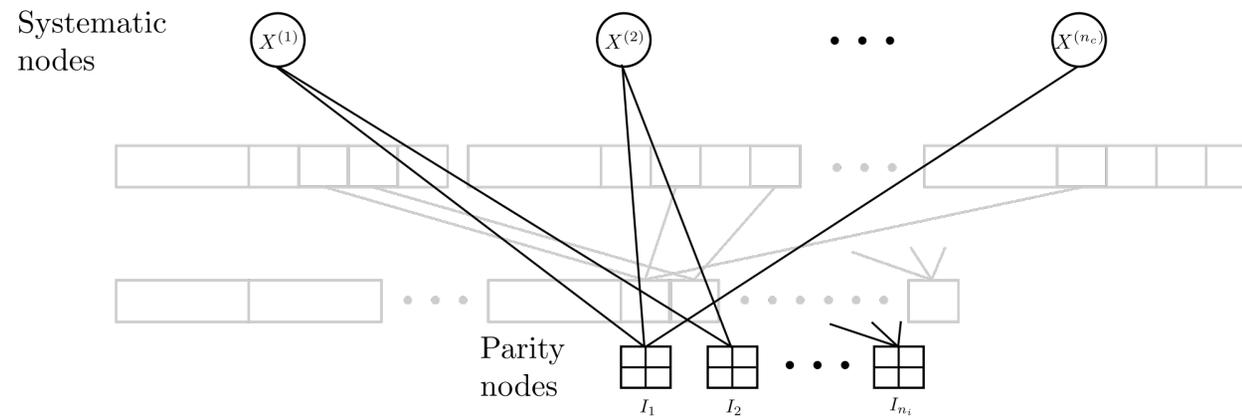
# Inter-frame Coding (IFC) [Zeineddine & Mansour]



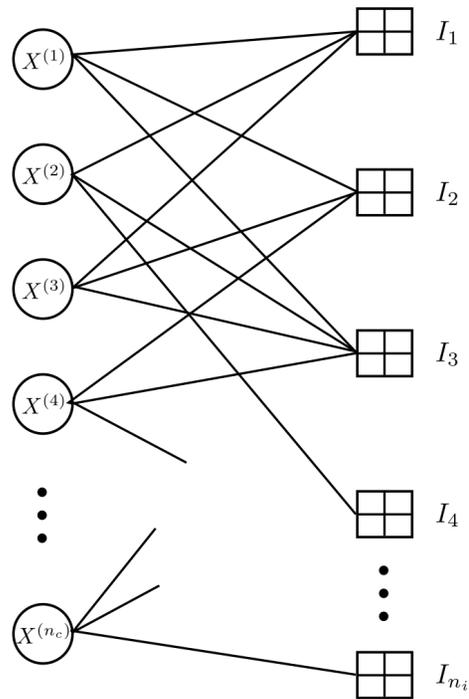
# IFC as a Low-Density Generator Matrix



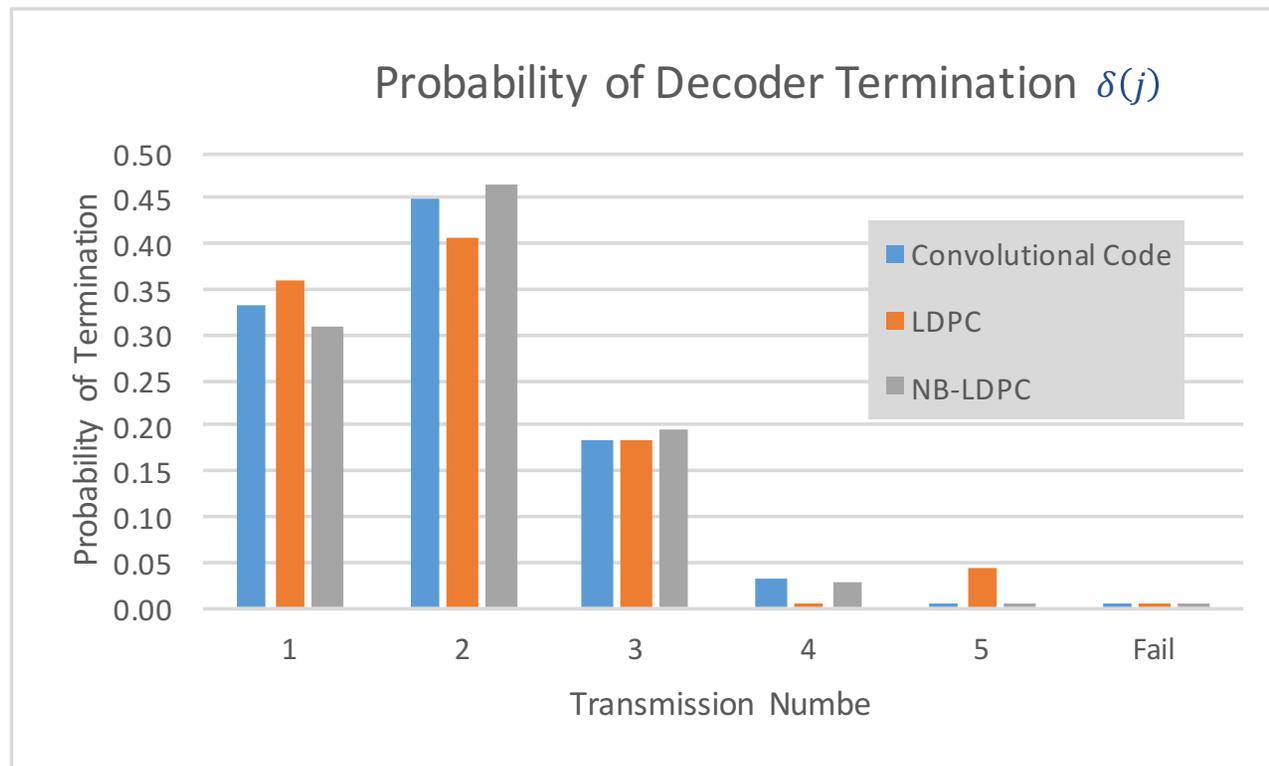
# IFC as a Low-Density Generator Matrix



# Generalized Peeling Decoder for IFC

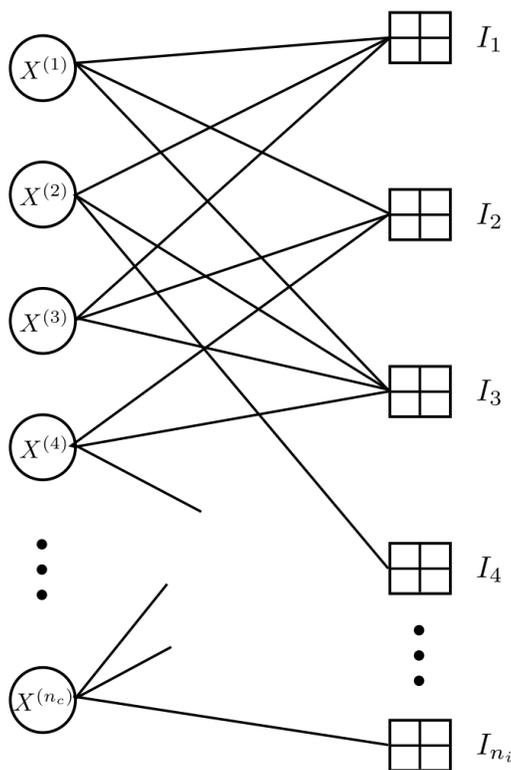


# Probability of Termination for our codes



# What degree distributions allow GPD to converge?

$$\lambda(x) = \sum_{i=1}^{d_L} \lambda_i x^{i-1}$$

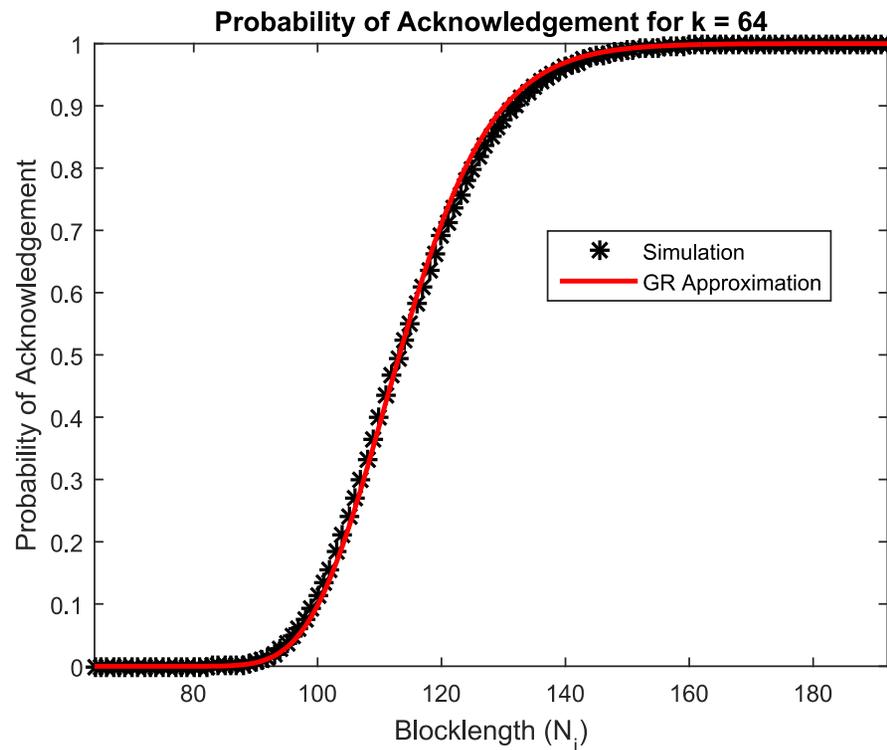


$$\rho(x) = \sum_{i=1}^{d_R} \rho_i x^{i-1}$$

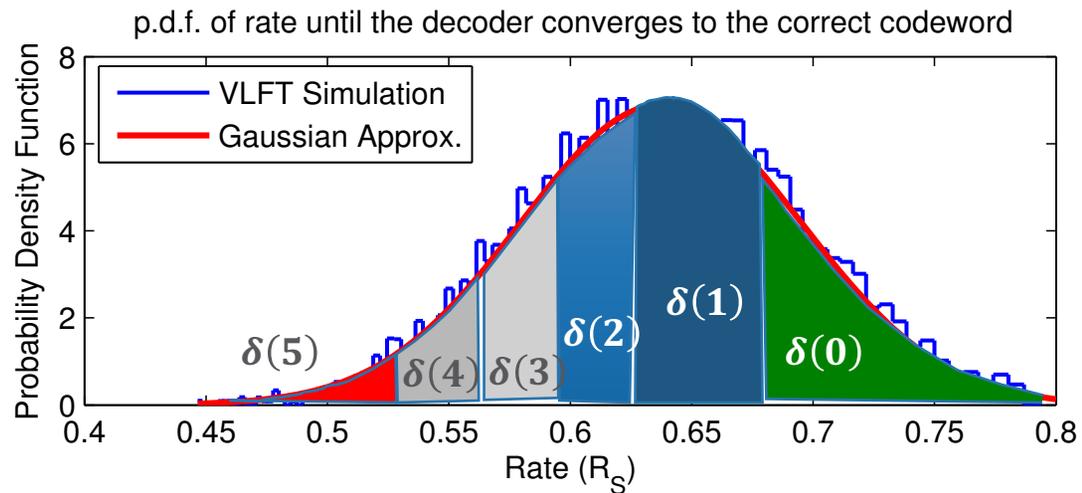
## What degree distributions allow GPD to converge?

- Following [Luby *et al.* Trans. IT 2001] and assuming a **geometric PMF** for  $\delta(j)$  [Zeineddine & Mansour] found a sequence of  $\lambda(x)$  and  $\rho(x)$  distributions as parameters  $d \rightarrow \infty$  and  $j \rightarrow \infty$ .
- But these distributions have **unbounded support** and anyway  $\delta(j)$  turns out to follow a **Gaussian model**.

# Rate of first decoding is well-modeled by Gaussian.



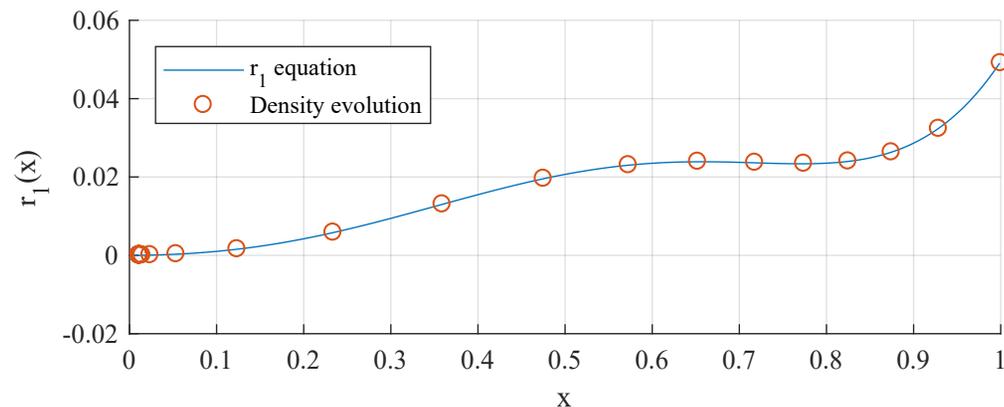
# $\delta(j)$ PMF as slices from the Gaussian



## Generalizing Luby's equation

$$\ell(x) = \sum_{\omega=1}^m \delta_{\omega} \sum_{i=1}^{d_L} \lambda_i \sum_{j=0}^{\min(\omega, i)-1} \binom{i-1}{j} (1-x)^j x^{i-1-j}.$$

$$r_1(x) = \ell(x)[\rho(1 - \ell(x)) - (1 - x)],$$



## The probability of VL code not decoding

$$\epsilon_{GPD}(i, \omega) = \sum_{j=0}^{\min(\omega-1, i)} \binom{i}{j} (1-x_0)^j x_0^{i-j}. \quad (12)$$

As a result, the probability of failure can be calculated as

$$\epsilon_{GPD} = \sum_{\omega=1}^m \delta_{\omega} \sum_{i=1}^{d_L} \Lambda_i \sum_{j=0}^{\min(\omega-1, i)} \binom{i}{j} (1-x_0)^j x_0^{i-j} \quad (13)$$

$$= \sum_{i=1}^{d_L} \Lambda_i \sum_{j=0}^{\min(m-1, i)} \gamma_j \binom{i}{j} (1-x_0)^j x_0^{i-j}, \quad (14)$$

where  $\gamma_j = \sum_{\omega=j+1}^m \delta_{\omega}$  as in  $\ell(x)$ , and  $\Lambda_i = \frac{\lambda_i/i}{\sum_{j=1}^{d_L} \lambda_j/i}$  is the left *node* degree distribution.



# Great performance with almost-regular $\rho(x)$

TABLE II  
PERFORMANCE CHARACTERISTICS OF CONCENTRATED  
 $\rho(x) = \alpha x^2 + (1 - \alpha)x^3$ .  $\lambda(x) = x^3$  IN ALL CASES.

$\alpha$	$a_R$	$\beta$	iter.	% $R_t^{(FB)}$	$\epsilon_{FF}$
1	3	1.333333	15	93.52%	$7.09 \times 10^{-4}$
0.531	3.39847	1.17700	20	95.36%	$7.82 \times 10^{-4}$
0.244	3.69914	1.08133	30	96.52%	$8.35 \times 10^{-4}$
0.168	3.78788	1.05600	40	96.83%	$8.50 \times 10^{-4}$
0.139	3.82287	1.04633	50	96.94%	$8.56 \times 10^{-4}$
0.108	3.86100	1.03600	100	97.08%	$8.63 \times 10^{-4}$

Channel Code Analysis and Design using Multiple Variable-Length Codes in Parallel without Feedback

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**Abstract**—This paper considers a channel coding paradigm that enables high throughput by using many variable-length codes in parallel, where each of the parallel codes has a short average blocklength. The inter-frame coding of Zaineddine and Mansour provides variable-length codes with incremental redundancy from a common pool of redundancy in a way that does not require feedback. A probability-based derivation of a generalized peeling decoder extends the results of Luby *et al.* to the inter-frame decoder in the inter-frame system. A new expression characterizes the probability that a variable-length decoder for throughput loss as compared to the original feedback system are identified, yielding a new, and far simpler, quasi-regular design methodology for the right degree distribution of the inter-frame code. The inter-frame paradigm can apply to any communication channel, but this paper uses the additive white Gaussian noise channel to demonstrate the concepts.

## I. INTRODUCTION

Practical systems and theoretical analysis [1]–[3] show that using a variable-length (VL) code with incremental transmissions controlled by ACK/NACK feedback can approach capacity with short average blocklengths on the order of 200–500 symbols. This paper studies the analysis and design of systems that use many variable-length codes in parallel and *without feedback* to approach capacity for point-to-point communication.

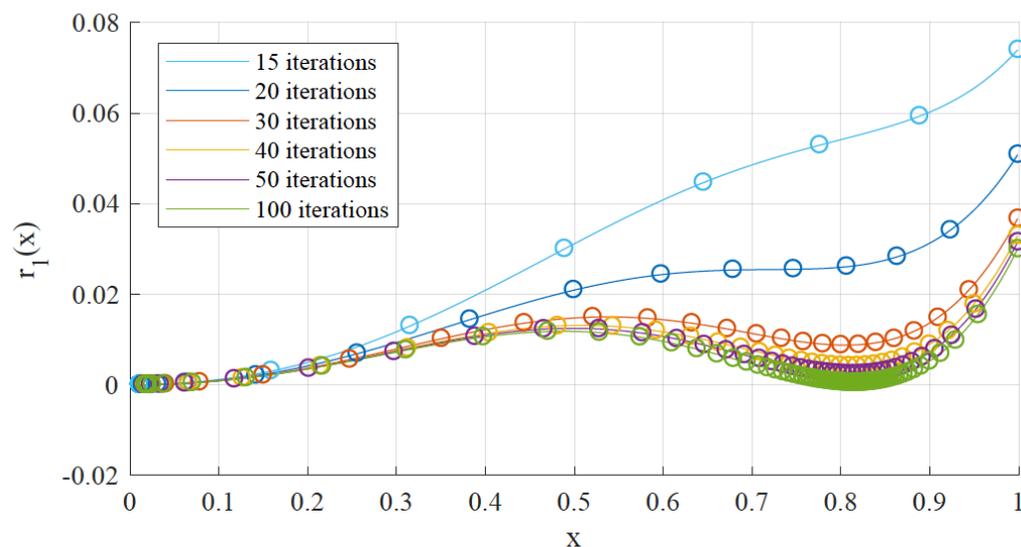
As described in [4] a large number of capacity-approaching VL codes can be decoded in parallel without feedback using the inter-frame coding approach of Zaineddine and Mansour [5], where an appropriate number of linear combinations of incremental redundancy (CPR), are transmitted. This can also be generalized to a doubly generalized LDPC (DGLDPC) [5], where an appropriate number of CPR provides incremental redundancy applied to the CPR decoders. Peeling decoders as an example of VL decoders. Peeling decoders for multiple access channels in parallel without feedback can be achieved.

The parallel system has as its conceptual VL code, which for the examples in this paper is a convolutional code with the rebit decoding algorithm in [4] and pseudo-random number of increments required (feedback) determines an upper bound on the expected number of increments required for a proposed system (that does not use feedback) to correct for the code failure; the very low frame error rates required for applications can be achieved.

The inter-frame code that generalizes the parallel system is described by a VL code, which for the examples in this paper is a convolutional code with the rebit decoding algorithm in [4] and pseudo-random number of increments required (feedback) determines an upper bound on the expected number of increments required for a proposed system (that does not use feedback) to correct for the code failure; the very low frame error rates required for applications can be achieved.

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# Great performance with almost-regular $\rho(x)$



Haobo Wang

Fig. 5.  $r_1(x)$  vs.  $x$  for Table II. The curves are generated using (10). Circles indicate iteration points determined through density evolution. The circles at  $x = 1$  represent the first iteration.

# Simulations compared to Density Evolution

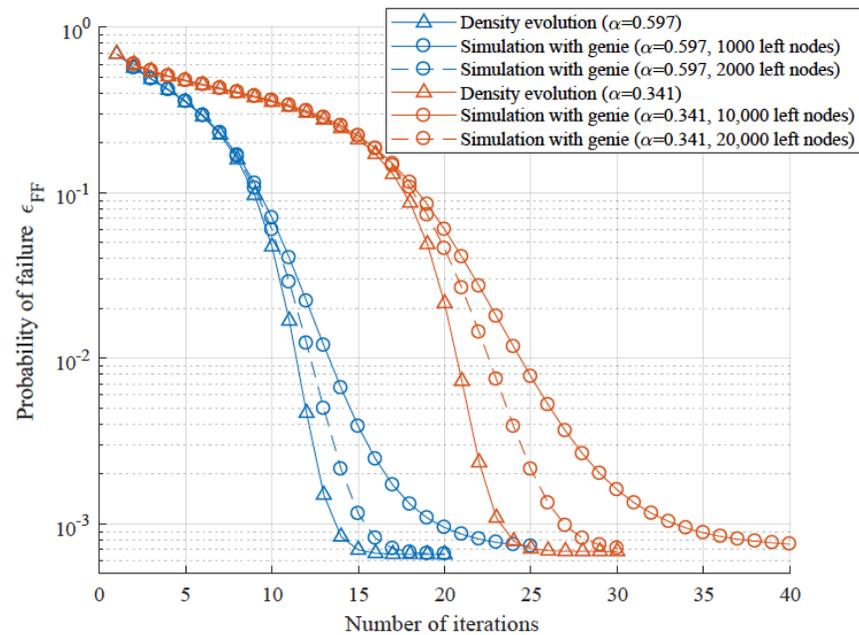


Fig. 6. Probability of failure vs. the number of iterations for the designs in Table III from density evolution and genie-aided simulations.

## Rate loss due to extra right nodes

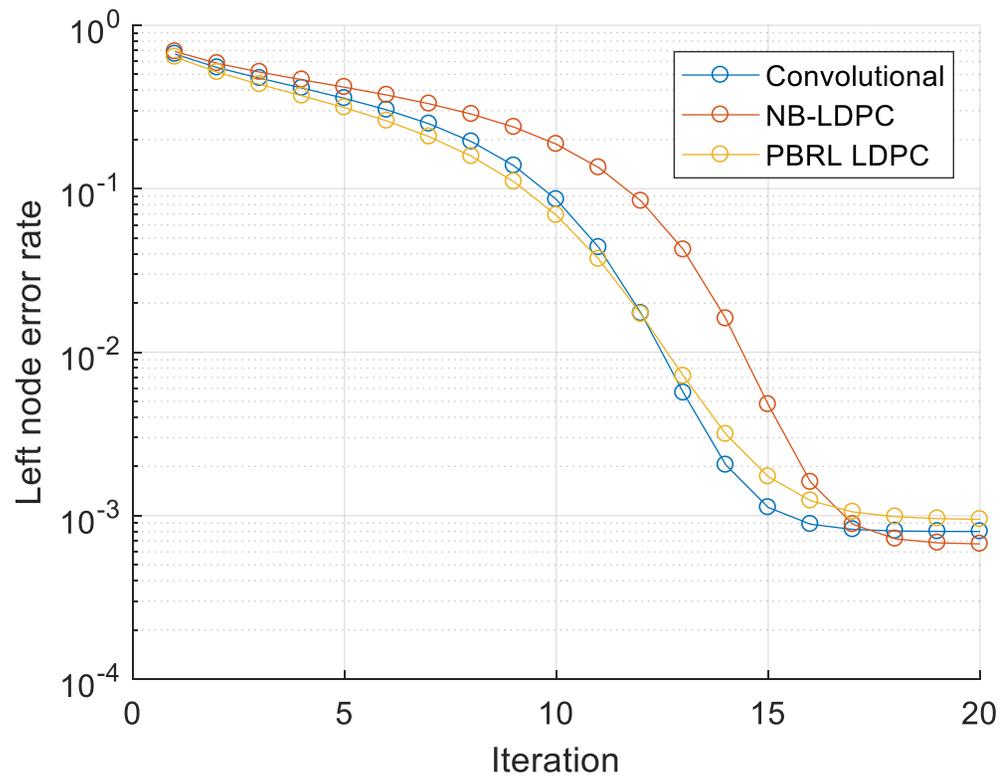
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$$R_t^{(FB)} = \frac{k(1 - \epsilon_{FB})}{\ell_0 + \beta_{FB}\ell_{\Delta}}$$

$$R_t^{(FF)} = \frac{k(1 - \epsilon_{FF})}{\ell_0 + \beta_{FF}\ell_{\Delta}}$$

$$\beta_{FF} = \frac{d_L}{a_r} = \frac{4}{3 + (1 - \alpha)}$$

# Density Evolution for Peeling Decoders



	Convolutional Code	LDPC	NB-LDPC
alpha	0.432	0.43	0.466
ar	3.568	3.57	3.534
Bff	1.12	1.12	1.13
Rff	0.507681527	0.55001249	0.55961973
Percent of FB	97.42%	98.57%	98.04%

# Engineer Change.