

# Layered Constructions for Low-Delay Streaming Codes

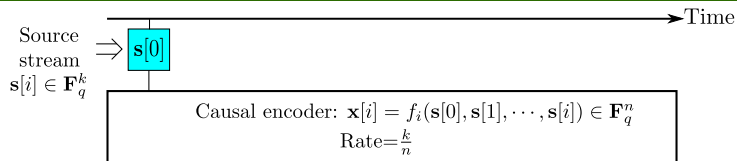
Ashish Khisti  
University of Toronto

Joint Work:  
Ahmed Badr (Toronto),  
Wai-Tian Tan (Cisco),  
John Apostolopoulos (Cisco).

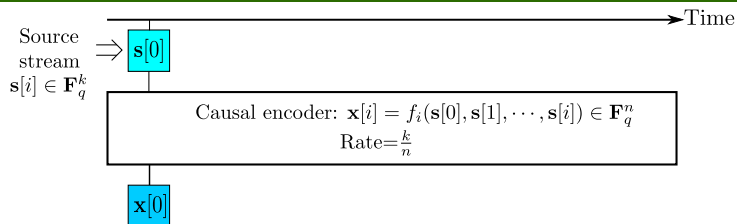
ITA, 2014

# Real-Time Communication System

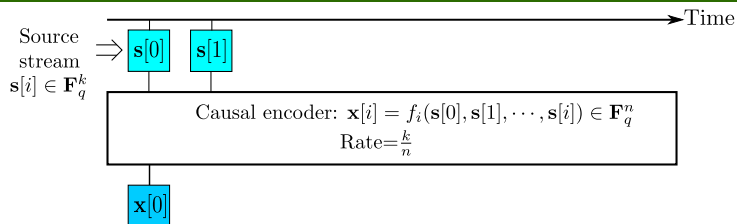
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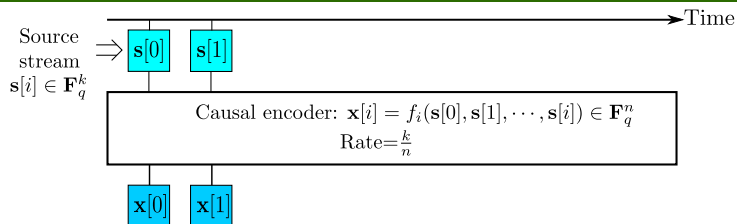
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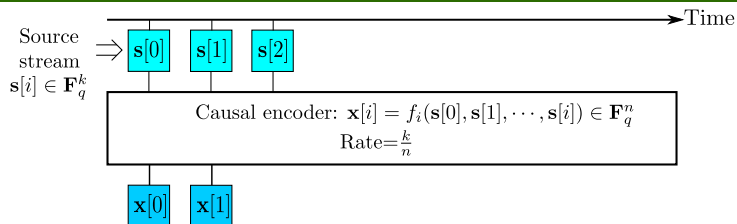
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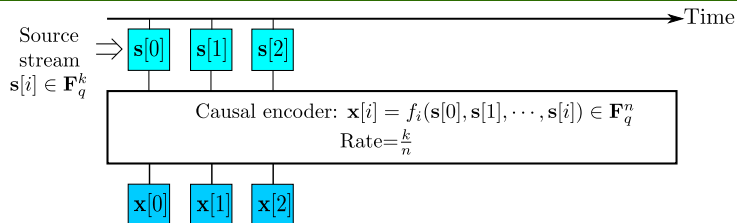
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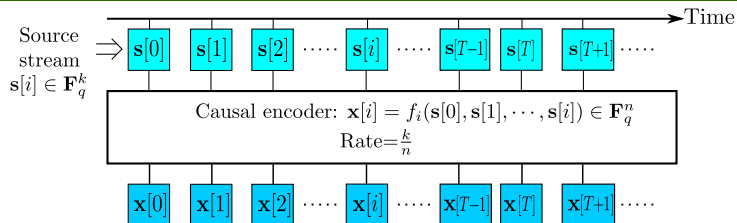


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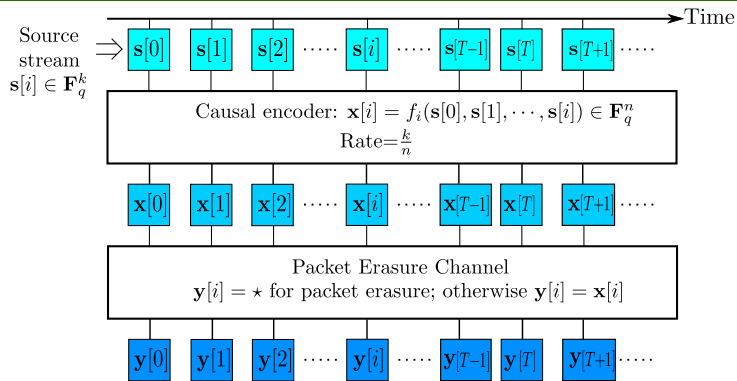




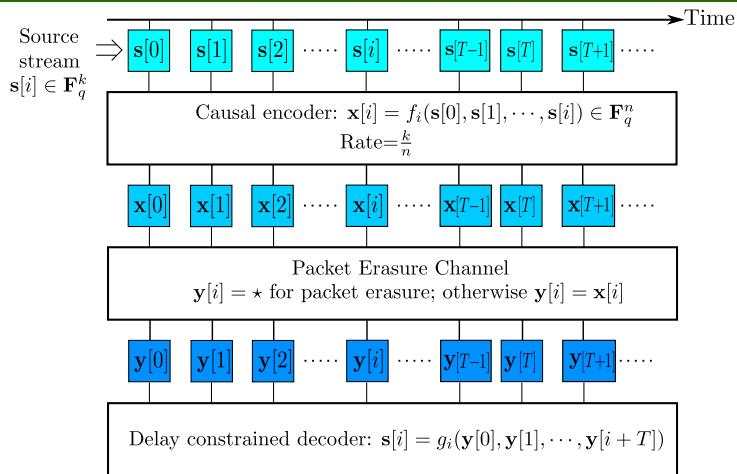
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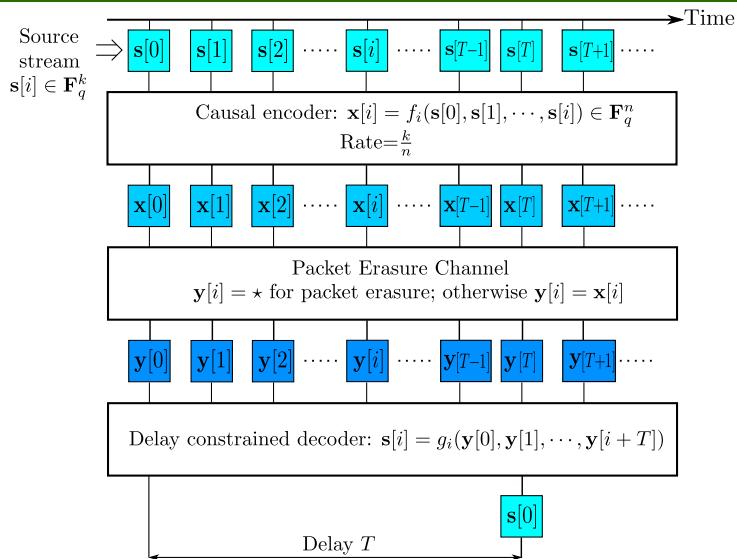
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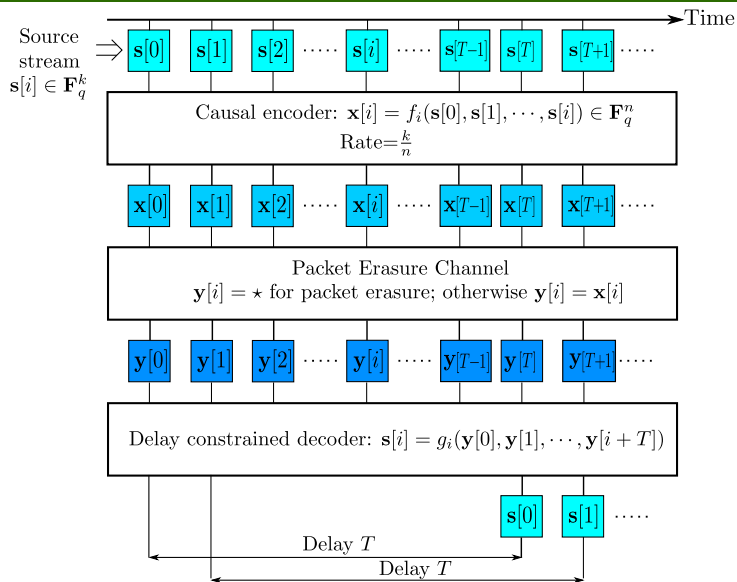
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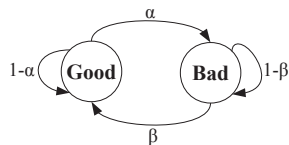


# Problem Setup

- **Source Model** : i.i.d. sequence  $s[t] \sim p_s(\cdot) = \text{Unif}\{(\mathbb{F}_q)^k\}$
- **Streaming Encoder**:  $x[t] = f_t(s[1], \dots, s[t])$ ,  $x[t] \in (\mathbb{F}_q)^n$
- Erasure Channel (To be specified)
- **Delay-Constrained Decoder**:  $\hat{s}[t] = g_t(y[1], \dots, y[t+T])$
- Rate  $R = \frac{k}{n}$

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## Gilbert-Elliott Model

- How much performance gains can we obtain?
- What are the fundamental metrics for low-delay error correction codes?

- **Structural Results on Real Time Coding:** Witsenhausen (1979), Teneketzis (2006), Mahajan (2009), Kaspı and Merhav (2012), Asnani and Weissman (2013) ...
- **Delay-Universal Codes:** Schulman (IT 1996), Sahai (2001), Martinian and Wornell (Allerton 2004), Sukhavasi and Hassibi (2011)
- **Low-Delay Codes:** Martinian and Sundberg (2004), Martinian (2004)

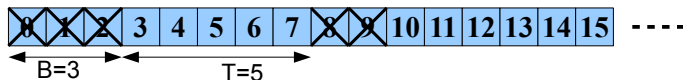


# Streaming Codes - Burst Erasure Channel

Martinian and Sundberg (2004)

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**B=3, T=5**



Capacity  $C(B, T)$

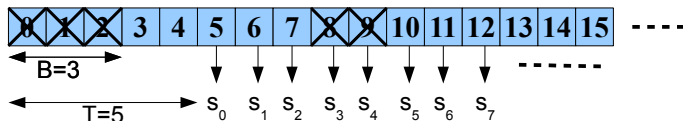
The maximum rate  $R$  such that there exists a rate  $R = \frac{k}{n}$  streaming code over a sufficiently large field  $q$  such that  $\Pr(s[t] \neq \hat{s}[t]) = 0$ , for all  $t \geq 0$ .

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## Capacity Result

Streaming Capacity:

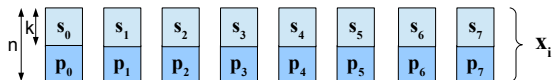
$$C(B, T) = \begin{cases} \frac{T}{T+B}, & T \geq B \\ 0, & T < B. \end{cases}$$

## Code Construction

MS Codes (Maximally-Short Codes) Construction:

- Step 1: Construct a low-delay block code
- Step 2: Use diagonal interleaving to construct a streaming code.

# Baseline Codes



$$\mathbf{p}_i = \mathbf{s}_i \cdot \mathbf{H}_0 + \mathbf{s}_{i-1} \cdot \mathbf{H}_1 + \dots + \mathbf{s}_{i-M} \cdot \mathbf{H}_M, \quad \mathbf{H}_i \in \mathbb{F}_q^{k \times n-k}$$

Erasure Codes:

- Random Linear Codes
- Strongly-MDS Codes (Gabidulin'88, Gluesing-Luerssen'06 )

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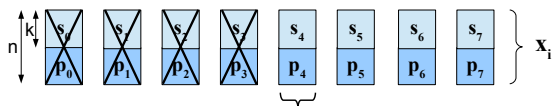


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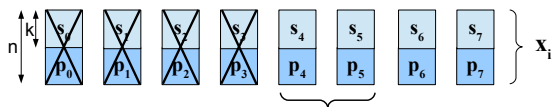


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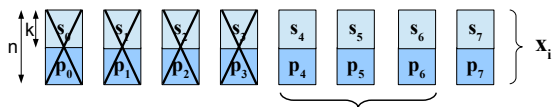


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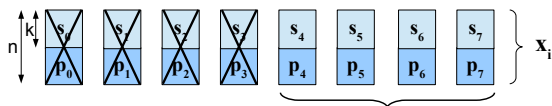
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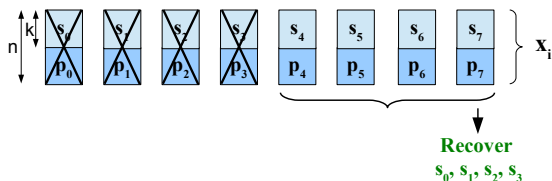


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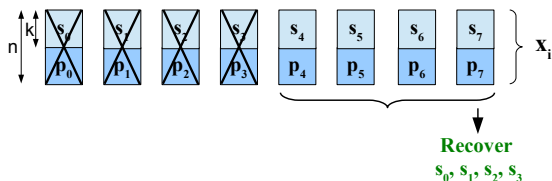


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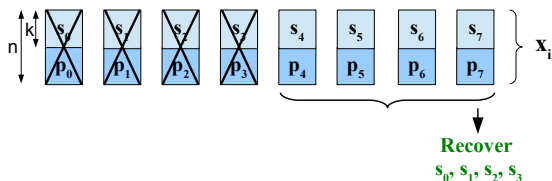


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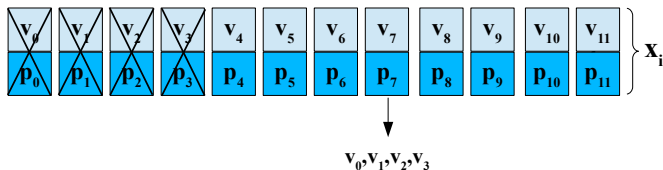
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$$\begin{bmatrix} \mathbf{p}_4 \\ \mathbf{p}_5 \\ \mathbf{p}_6 \\ \mathbf{p}_7 \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{H}_4 & \mathbf{H}_3 & \mathbf{H}_2 & \mathbf{H}_1 \\ \mathbf{H}_5 & \mathbf{H}_4 & \mathbf{H}_3 & \mathbf{H}_2 \\ 0 & \mathbf{H}_5 & \mathbf{H}_4 & \mathbf{H}_3 \\ 0 & 0 & \mathbf{H}_5 & \mathbf{H}_4 \end{bmatrix}}_{\text{full rank}} \begin{bmatrix} \mathbf{s}_0 \\ \mathbf{s}_1 \\ \mathbf{s}_2 \\ \mathbf{s}_3 \end{bmatrix}$$

# Streaming Code - Example

$$B = 4, T = 8$$

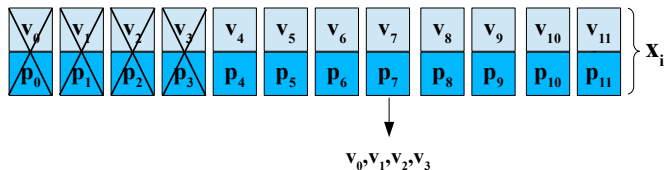
Rate 1/2 Baseline Erasure Codes,  $T = 7$



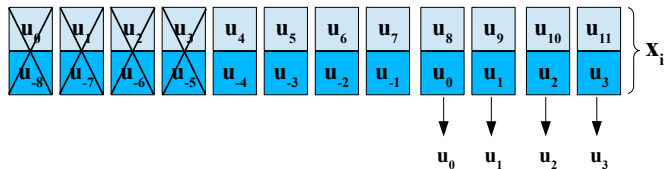
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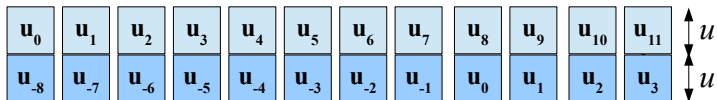
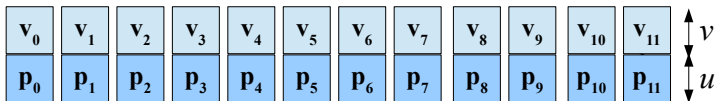


Rate 1/2 Repetition Code,  $T = 8$



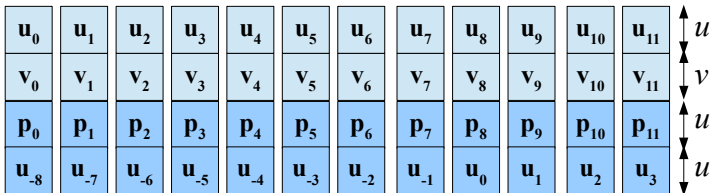
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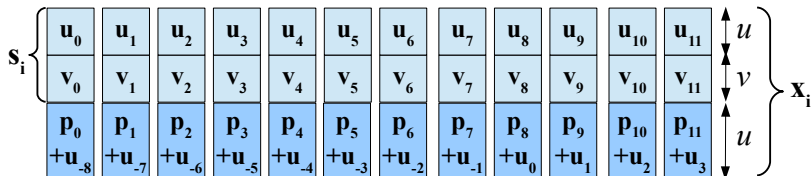


$$R = \frac{u+v}{3u+v} = \frac{1}{2}$$



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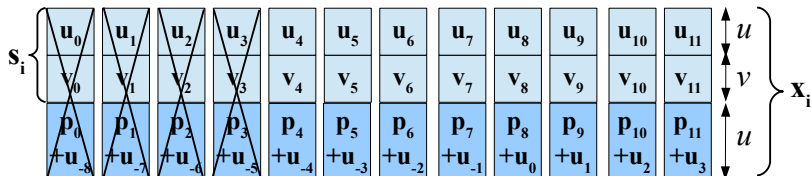
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$$R = \frac{u+v}{2u+v} = \frac{2}{3}$$

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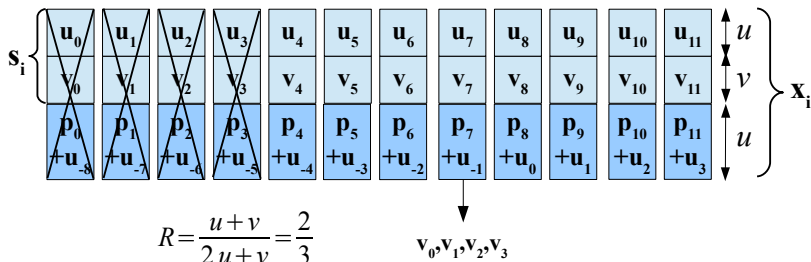
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## Encoding:

- 1 Split each source symbol into 2 groups  $s_i = (u_i, v_i)$
- 2 Apply Erasure code to the  $v_i$  stream generating  $p_i$  parities
- 3 Repeat the  $u_i$  symbols with a shift of  $T$
- 4 Combine the repeated  $u_i$ 's with the  $p_i$ 's

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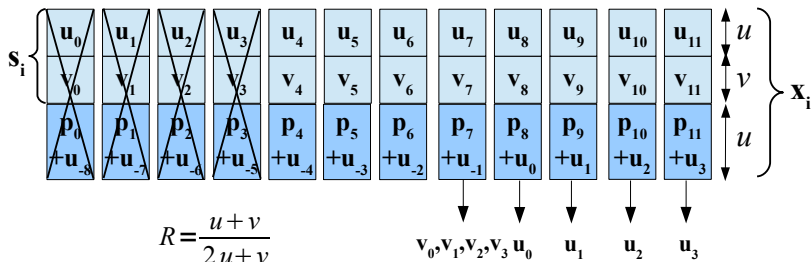


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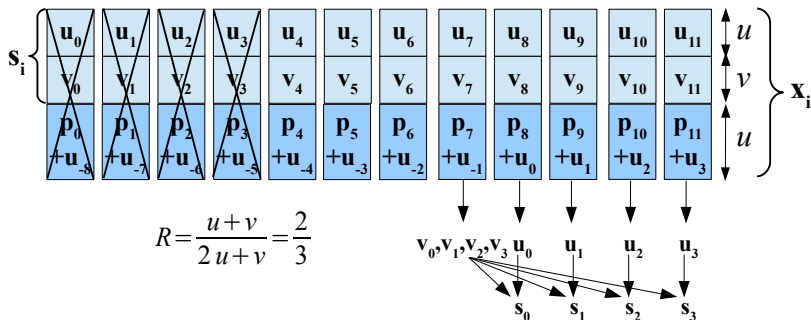
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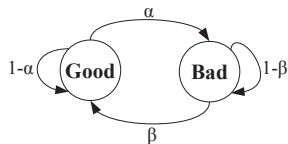
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# Simulation

Gilbert-Elliott Channel  $(\alpha, \beta) = (5 \times 10^{-4}, 0.5)$ ,  $T = 12$  and  $R = 12/23$

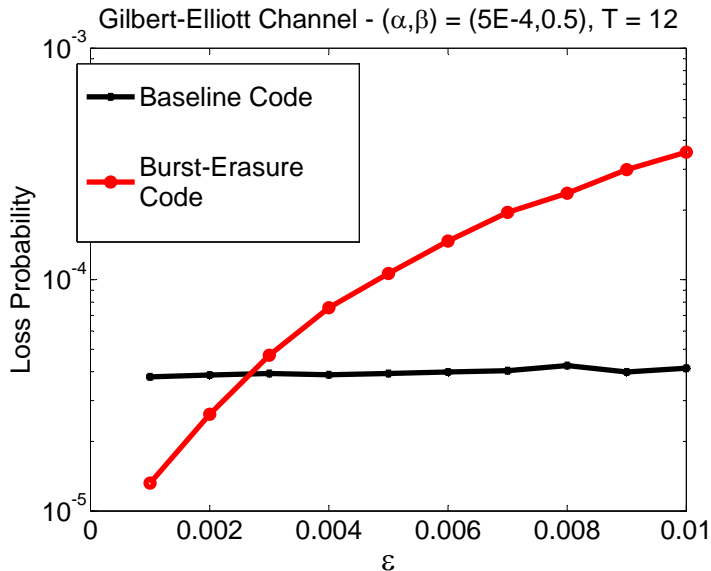
## Gilbert Elliott Channel

- Good State:  $\Pr(\text{loss}) = \varepsilon$
- Bad State:  $\Pr(\text{loss}) = 1$



# Simulation

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- Rate  $R = \frac{k}{n}$

## Sliding Window Erasure Channel

In any sliding window of length  $W$ , the channel can introduce only one of the following:

- An erasure burst of maximum length  $B$
- Upto  $N$  erasures in arbitrary positions

# Problem Setup

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**$(N, B, W) = (2, 3, 6)$**

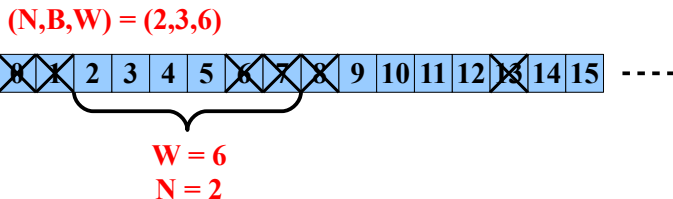


**$W = 6$**

**$N = 2$**

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**$(N, B, W) = (2, 3, 6)$**



**$W = 6$**

**$B = 3$**

Capacity:  $C(N, B, W, T)$

## Theorem

Consider the  $\mathcal{C}(N, B, W)$  channel, with  $W \geq B + 1$ , and let the delay be  $T$ .

**Upper-Bound** For any rate  $R$  code, we have:

$$\left(\frac{R}{1-R}\right) B + N \leq \min(W, T + 1)$$

**Lower-Bound:** There exists a rate  $R$  code that satisfies:

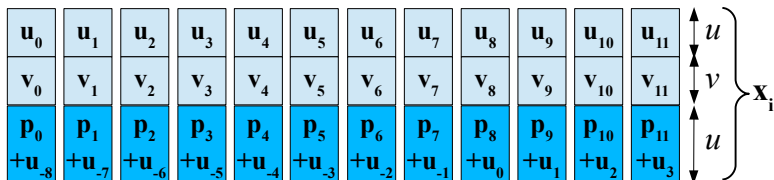
$$\left(\frac{R}{1-R}\right) B + N \geq \min(W, T + 1) - 1.$$

The gap between the upper and lower bound is 1 unit of delay.

# Streaming Codes - Isolated Erasures

$C(N \geq 2, B, W)$

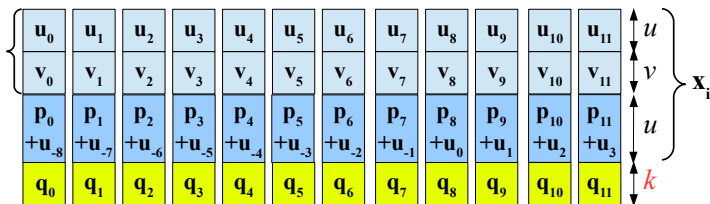
$$T = 8$$



- Erasures at time  $t = 0$  and  $t = 8$
- $\mathbf{u}_0$  cannot be recovered due to a repetition code

# Proposed Approach: Layering

$C(N \geq 2, B, W)$



## Layered Code Design

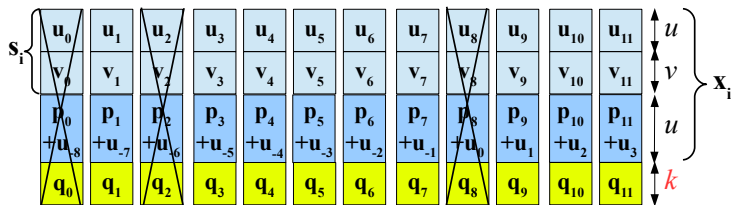
- Burst-Erasure Streaming Code  $\mathcal{C}_1 : (\mathbf{u}_i, \mathbf{v}_i, \mathbf{p}_i + \mathbf{u}_{i-T})$
- Erasure Code:  $\mathbf{q}_i = \sum_{t=1}^M \mathbf{u}_{i-t} \cdot \mathbf{H}_t^u$ ,  $\mathbf{q}_i \in \mathbb{F}_q^k$
- $\mathcal{C}_2 : (\mathbf{u}_i, \mathbf{v}_i, \mathbf{p}_i + \mathbf{u}_{i-T}, \mathbf{q}_i)$

$$R = \frac{T}{T + B + k}, \quad k = \frac{N}{T - N + 1}B$$



# Proposed Approach: Layering

$C(N \geq 2, B, W)$



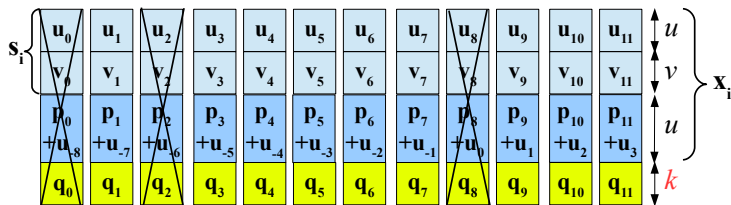
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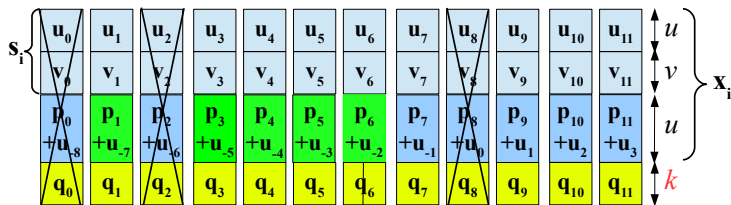
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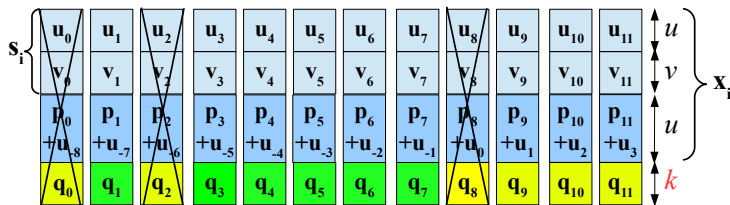
Layered Code Design

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$C(N \geq 2, B, W)$



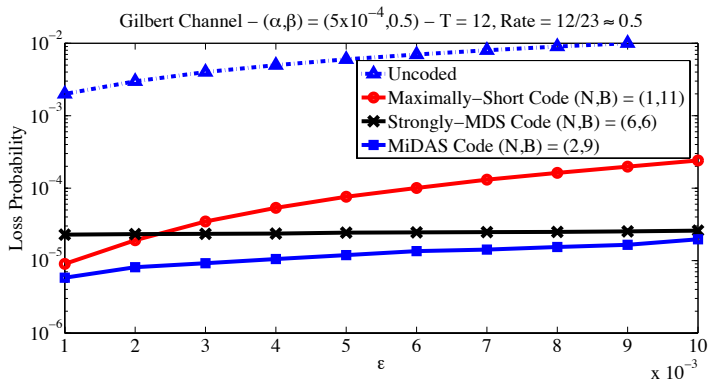
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# Simulation Results

Gilbert-Elliott Channel  $(\alpha, \beta) = (5 \times 10^{-4}, 0.5)$ ,  $T = 12$  and  $R = 12/23 \approx 0.5$

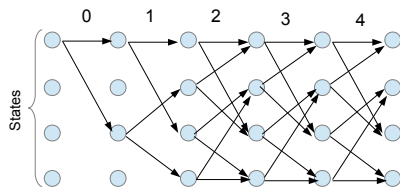


Code	N	B	Code	N	B
Strongly MDS	6	6	MiDAS	2	9
Burst-Erasure	1	11			

# Distance and Span Properties

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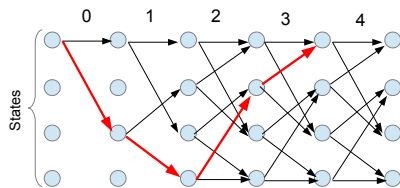
Consider  $(n, k, m)$  Convolutional code:  $\mathbf{x}_i = \sum_{j=0}^m \mathbf{s}_{i-j} \mathbf{G}_j$



Trellis Diagram

# Distance and Span Properties

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Trellis Diagram – Free Distance



# Distance and Span Properties

Consider  $(n, k, m)$  Convolutional code:  $\mathbf{x}_i = \sum_{j=0}^m \mathbf{s}_{i-j} \mathbf{G}_j$

Column Distance:  $d_T$

$$d_T = \min_{\substack{[\mathbf{s}_0, \dots, \mathbf{s}_T] \\ \mathbf{s}_0 \neq 0}} \text{wt} \left( \begin{bmatrix} \mathbf{s}_0 & \dots & \mathbf{s}_T \end{bmatrix} \begin{bmatrix} \mathbf{G}_0 & \mathbf{G}_1 & \dots & \mathbf{G}_T \\ 0 & \mathbf{G}_0 & \dots & \mathbf{G}_{T-1} \\ \vdots & & \ddots & \vdots \\ 0 & & \dots & \mathbf{G}_0 \end{bmatrix} \right)$$

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Column Span:  $c_T$

$$c_T = \min_{\substack{[\mathbf{s}_0, \dots, \mathbf{s}_T] \\ \mathbf{s}_0 \neq 0}} \text{span} \left( \begin{bmatrix} \mathbf{s}_0 & \dots & \mathbf{s}_T \end{bmatrix} \begin{bmatrix} \mathbf{G}_0 & \mathbf{G}_1 & \dots & \mathbf{G}_T \\ 0 & \mathbf{G}_0 & \dots & \mathbf{G}_{T-1} \\ \vdots & & \ddots & \vdots \\ 0 & & \dots & \mathbf{G}_0 \end{bmatrix} \right)$$

# Column-Distance & Column Span Tradeoff

## Theorem

*Consider a  $\mathcal{C}(N, B, W)$  channel with delay  $T$  and  $W \geq T + 1$ . A streaming code is feasible over this channel if and only if it satisfies:  $d_T \geq N + 1$  and  $c_T \geq B + 1$*

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## Theorem

For any rate  $R$  convolutional code and any  $T \geq 0$  the Column-Distance  $d_T$  and Column-Span  $c_T$  satisfy the following:

$$\left( \frac{R}{1-R} \right) c_T + d_T \leq T + 1 + \frac{1}{1-R}$$

There exists a rate  $R$  code (MiDAS Code) over a sufficiently large field that satisfies:

$$\left( \frac{R}{1-R} \right) c_T + d_T \geq T + \frac{1}{1-R}$$

- **Burst plus Isolated Erasures:** Layered Coding Approach (Badr-K-Tan-Apostolopoulos 2013, Badr 2014)
- **Mismatched Source-Channel Rates:** (Patil-Badr-K-Tan 2013), (Leong-Ho 2012), (Leong-Qureshi-Ho 2013)
- **Multicast Streaming Codes:** Optimal Codes for certain parameters; Column-Span profile (Badr-Lui-K 2014)
- **Parallel Channels:** Lui-Badr-Khisti 2011
- **Multi-Source Models:** Lui 2011

## Streaming Codes for Real-Time Streaming over Channels with Burst and Isolated Erasures

- Sliding Window Erasure Channel Model
- MiDAS Codes: Near Optimal Distance/Span Tradeoff
- Layering Approach
- Distance and Span Metrics

## Future Work

- Improved constructions for short-inter burst gaps
- Systems Theoretic Approach (e.g. Dual Codes for MiDAS Codes)
- Analysis of probabilistic channels