

On the Capacity of Multiplicative Finite-Field Matrix Channels

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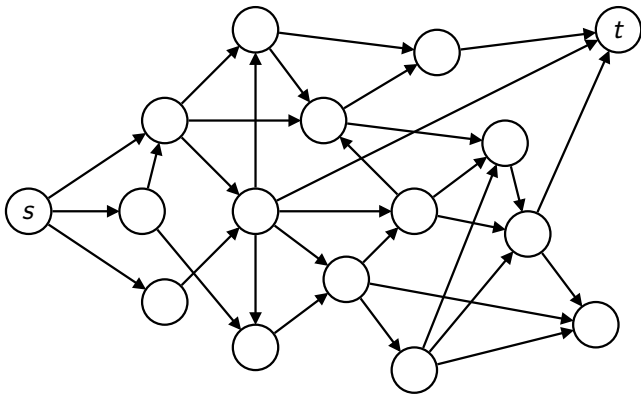
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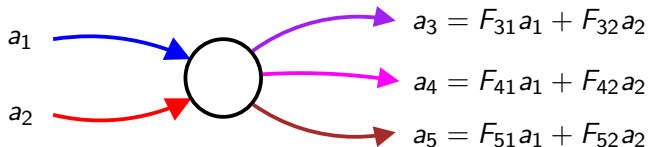
Motivation

Linear network coding



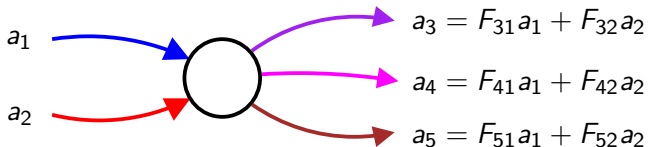
Linear network coding

- On each node: output is a **linear combination** of the input



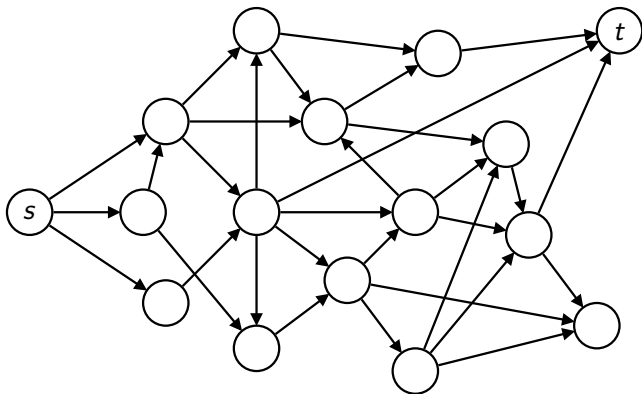
Linear network coding

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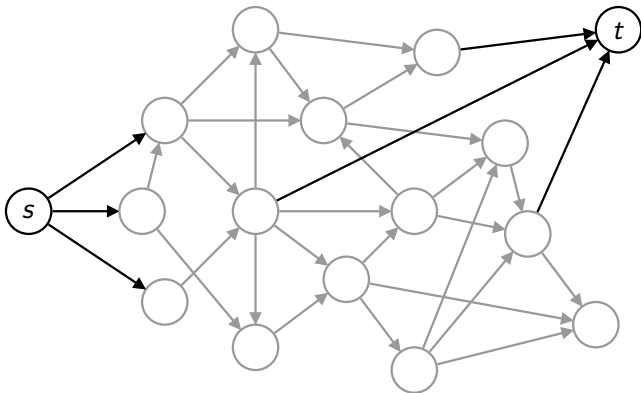


- Alphabet: **finite field** \mathbb{F}_q

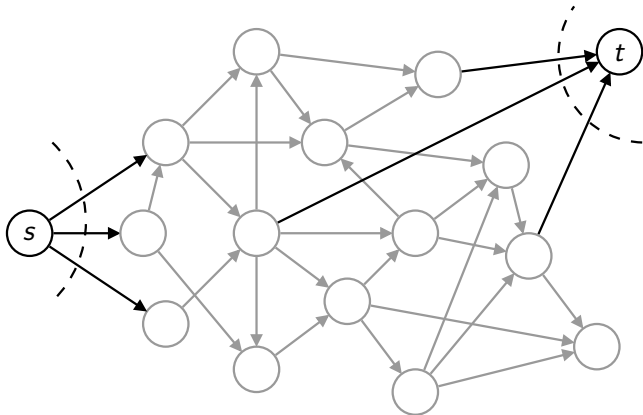
End-to-end approach



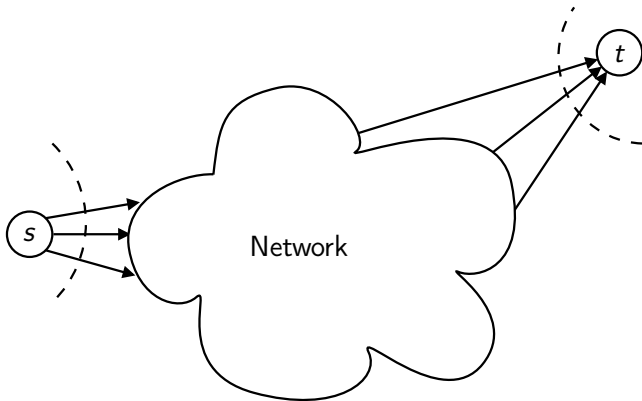
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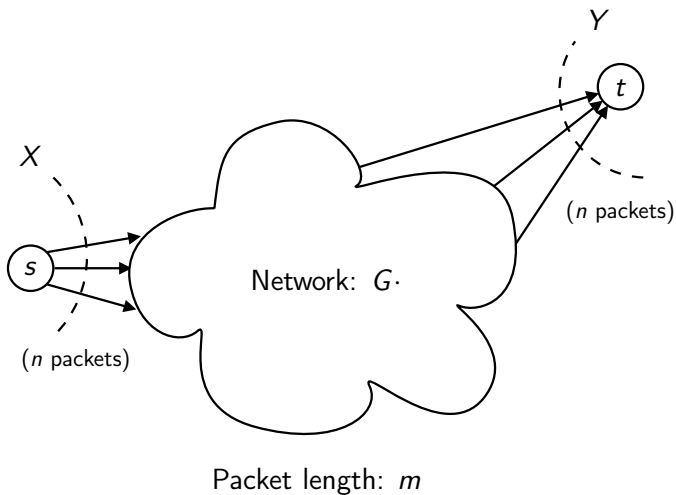
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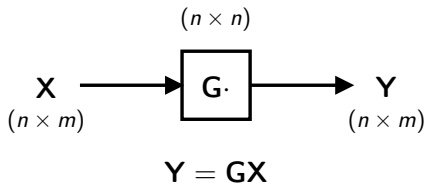
Channel model

Multiplicative finite-field matrix channels

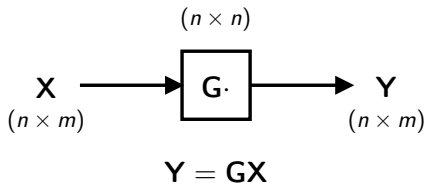


$$Y = GX$$

Multiplicative finite-field matrix channels

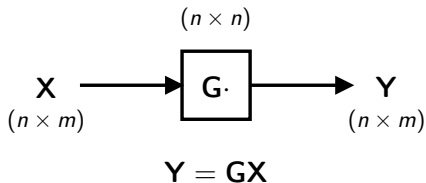


Multiplicative finite-field matrix channels



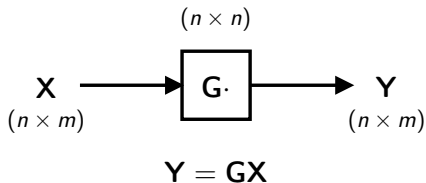
- Probabilistic model: matrices are **random variables** (**bold**)

Multiplicative finite-field matrix channels



- Probabilistic model: matrices are **random variables**
- **DMC** defined by $(\mathcal{X}, p(Y|X), \mathcal{Y})$

Multiplicative finite-field matrix channels



- Probabilistic model: matrices are **random variables**
- **DMC** defined by $(\mathcal{X}, p(Y|X), \mathcal{Y})$
- $p(Y|X)$ induced by $p(G)$ through the **channel law**:

$$p(Y|X) = \sum_G p(G) 1[Y = GX]$$

Previous works



$$Y = GX$$

- \mathbf{G} full-rank, uniform [Silva et al., 2010]
- \mathbf{G} with uniform i.i.d. entries \equiv \mathbf{G} uniform [Jafari et al., 2011]

Previous works



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 - **Too general**: complex channel description (q^{nm})

This work



$$Y = GX$$

- G uniform given rank (u.g.r.) \triangleq matrices with same rank are equiprobable

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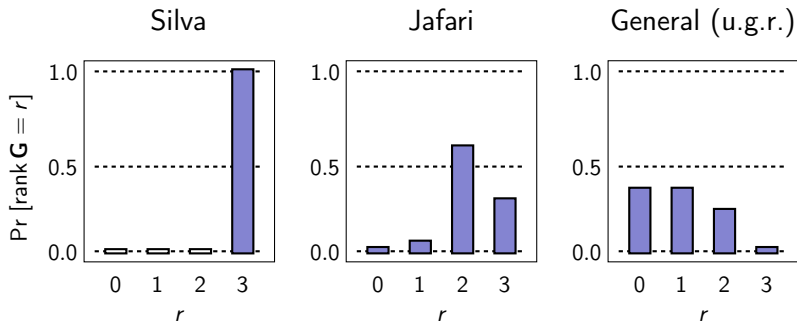
$$Y = GX$$

- **G** uniform given rank (u.g.r.) \triangleq matrices with same rank are equiprobable
 - Simple channel description ($n + 1$)
 - Keeps the essence of **non-coherence**
 - Serves a **lower bound** on the capacity for the general case:



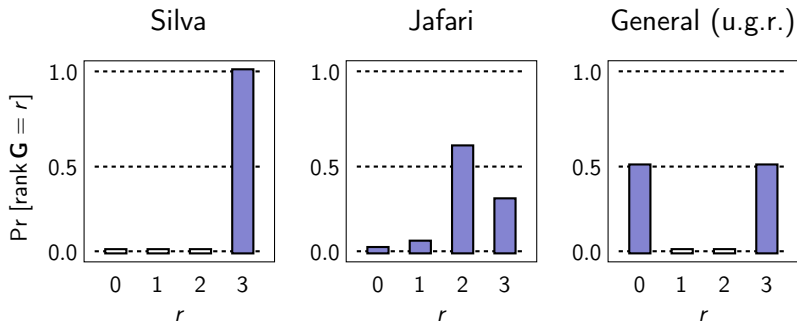
Comparison

- Example: \mathbf{G} of dimension 3×3 , binary field



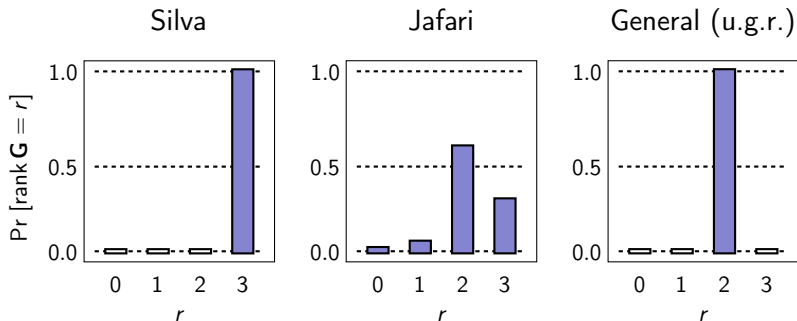
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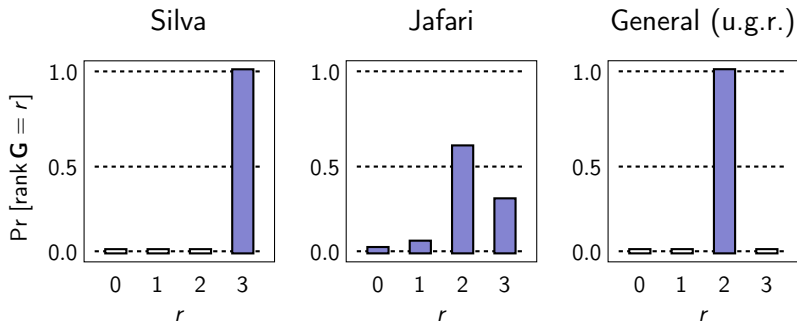
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Comparison

- Example: \mathbf{G} of dimension 3×3 , binary field



- Rank distribution depends on the **topology**, the **link erasure probabilities**, and the **linear combinations**

Our results

Channel transition probability

$$\mathbf{Y} = \mathbf{G}\mathbf{X}$$

$$\mathbf{u} \triangleq \text{rank } \mathbf{X}$$

$$\mathbf{v} \triangleq \text{rank } \mathbf{Y}$$

$$\mathbf{r} \triangleq \text{rank } \mathbf{G}$$

Channel transition probability

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Channel transition probability

$$p(Y|X) = \begin{cases} \frac{p(v|u)}{|\mathcal{T}_q^{n \times u, v}|}, & \text{if } \langle Y \rangle \subseteq \langle X \rangle, \\ 0, & \text{else.} \end{cases}$$

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- $p(v|u)$: rank transition probability:

$$p(v|u) = \sum_{r=0}^n p(r) \frac{|\mathcal{T}_q^{n \times u, v}|}{|\mathcal{T}_q^{n \times n, r}|} \phi_q(n; u, n, v, r),$$

calculated using a combinatorial result [Brawley and Carlitz, 1973]

Channel transition probability

$$\mathbf{Y} = \mathbf{GX}$$

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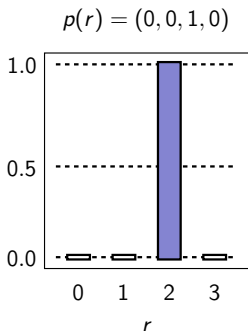
- $p(Y|X)$: $q^{nm} \times q^{nm}$ matrix
- $p(v|u)$: $(n+1) \times (n+1)$ matrix

Rank transition probability

$$\mathbf{Y} = \mathbf{GX}$$

$$\mathbf{u} \triangleq \text{rank } \mathbf{X} \quad \mathbf{v} \triangleq \text{rank } \mathbf{Y} \quad \mathbf{r} \triangleq \text{rank } \mathbf{G}$$

- Example: \mathbf{G} of dimension 3×3 , binary field



$$p(v|u) = \begin{array}{c|cccc} & \begin{array}{c} v \\ \hline u \end{array} & 0 & 1 & 2 & 3 \\ \hline 0 & & 1 & 0 & 0 & 0 \\ 1 & & \frac{1}{7} & \frac{6}{7} & 0 & 0 \\ 2 & & 0 & \frac{3}{7} & \frac{4}{7} & 0 \\ 3 & & 0 & 0 & 1 & 0 \end{array}$$

Channel capacity

- Maximization split in two stages:

$$\begin{aligned} C &= \max_{p(\mathbf{X})} I(\mathbf{X}; \mathbf{Y}) \\ &= \max_{p(\mathbf{u})} \max_{p(\mathbf{X}):p(\mathbf{u})} I(\mathbf{X}; \mathbf{Y}) \end{aligned}$$

Channel capacity

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Inner stage: Solved, max achieved with u.g.r. input

$$\max_{p(\mathbf{X}):p(u)} I(\mathbf{X}; \mathbf{Y}) = \sum_v p(v) \log_q \frac{a_v}{p(v)} - \sum_u b_u p(u) \triangleq I^*(p(u))$$

Channel capacity

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Outer stage: No closed-form solution

$$\max_{p(u)} I^*(p(u)) = \text{convex optimization problem}$$

Channel capacity

Original problem

$$C = \max_{p(X)} I(p(X))$$

Convex optimization on q^{nm} variables

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New problem

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Problems are equivalent

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Convex optimization on $n + 1$ variables

Upper bound

#1 – Actual optimization problem

$$C = \max_{p(u)} I^*(p(u))$$

subject to $p(0) + \dots + p(n) = 1 \quad \forall u : p(u) \geq 0$

Upper bound

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#2 – Modified optimization problem

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Problems are **not** equivalent

#2 – Modified optimization problem

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Upper bound

- Exact solution for #2:

$$\tilde{C} = \log_q \sum_{v=0}^n |\mathcal{T}_q^{n \times m, v}| q^{-c_v}$$

where c_v 's are the solution of a triangular linear system of equations

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- Upper bound: $C \leq \tilde{C}$
- Sometimes this is also the solution for #1!

Constant-rank input

- Input matrices limited to have rank u

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Rank- u capacity

$$C_u = \sum_{v=0}^n p(v|u) \log_q \frac{\begin{bmatrix} m \\ v \end{bmatrix}_q}{\begin{bmatrix} u \\ v \end{bmatrix}_q}$$

Constant-rank input

- Input matrices limited to have rank u

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- Asymptotically optimal: $\max_u C_u \leq C \leq \max_u C_u + \log_q n$

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Unconstrained capacity, for $q \rightarrow \infty$

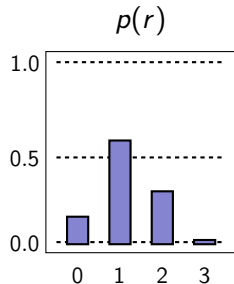
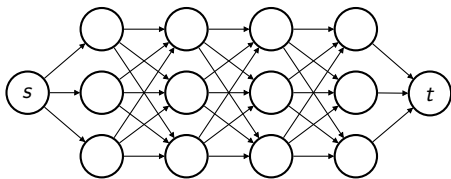
$$C = C_{u^*}$$

An example

“Trellis network”

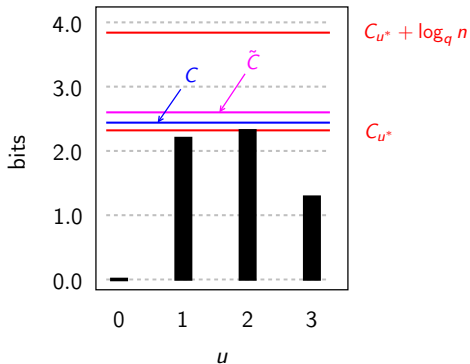
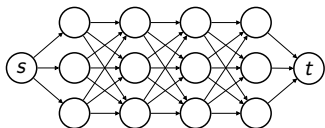
$\Pr[\text{erasure}] = 10\%$ for each link, binary field

$n = 3$ $m = 4$



An example

“Trellis network”



Communication via subspaces

- The matrix channel:

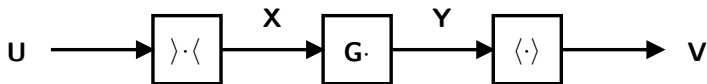


Communication via subspaces

- The matrix channel:



- The subspace channel:

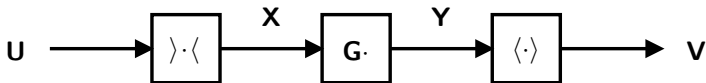


Communication via subspaces

- The **matrix channel**:



- The **subspace channel**:



- Communication via subspaces is **optimal** for G u.g.r.:

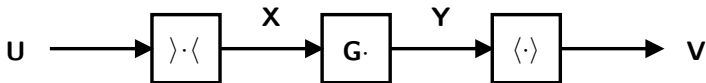
$$I(\mathbf{X}; \mathbf{Y}) = I(\mathbf{U}; \mathbf{V})$$

Communication via subspaces

- The **matrix channel**:



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- Communication via subspaces is **optimal** for G u.g.r.:

$$I(\mathbf{X}; \mathbf{Y}) = I(\mathbf{U}; \mathbf{V})$$

- Approach: “**grouping of letters**” in a DMC

Conclusion

Review

- U.g.r. transfer matrix: reasonable model for non-coherent networks

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- Capacity: optimization problem, bounds, special cases

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- U.g.r. transfer matrix: reasonable model for non-coherent networks
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- Communication via subspaces: still optimal

Review

- U.g.r. *transfer matrix*: reasonable model for non-coherent networks
- *Capacity*: optimization problem, bounds, special cases
- *Communication via subspaces*: still optimal
- Main open problem: *codes*

Thank you!

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