Common Information of Random Linear Network Coding Over A 1-Hop Broadcast Packet Erasure Channel

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Motivation — The COPE Principle

The COPE protocol — 2-hop relay networks [Katti et al. 06]

4 transmissions w/o coding vs. 3 transmissions w. coding

- r sends [X + Y]; d_1 decodes X by subtraction.
- Empirically, 40–200% throughput improvement.





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- r sends [X + Y]; d_1 decodes X by subtraction.
- Empirically, 40–200% throughput improvement.
- The capacity can be defined over the corresponding PEC network.







Assume round-based schemes, and $p_{r;d_1} \ge p_{r;d_2} > 0$.









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Can we combine the benefits of Network Coding & Opportunistic Routing?





Without direct $s_i \rightarrow d_i$ communication: $X_1 \cdots X_{nR_1} = Y_1 \cdots Y_{nR_2}$ 2-Stage Scheme takes the following as input:

> $p_{r;1}, p_{r;2}$: The CH parameters, nR_1, nR_2 : The total # of to-be-sent symbols, $np_{s_1;d_2}, np_{s_2;d_1}$: The overheard info.





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Wang, ISIT 2011 – p. 5/15



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> > Wang, ISIT 2011 – p. 5/15

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• Without direct $s_i \rightarrow d_i$ communication:

With this motivation, this work studies the Common Info. of Random Linear Network Coding.

• With direct $s_i \rightarrow d_i$ communication: We thus need

- Random Linear Network Coding (RLNC)
 - N packets: $\mathbf{W} \stackrel{\Delta}{=} (W_1, \cdots, W_N) \in (\mathsf{GF}(q))^N$.
 - Each time t, source sends $Y_t = \mathbf{v}_t \mathbf{W}^T$ via an erasure CH.
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 - Knowledge Space: $\Omega_k = \text{span}\{\mathbf{v}_t : \forall t \text{ such that } d_k \in \mathcal{R}_t\}.$
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- Common Information (CI): rank $\left(\bigcap_{k=1}^{K} \Omega_{k}\right)$.
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• Other notations: $\mathcal{T}_S \triangleq |\{\forall t : S \subseteq \mathcal{R}_t|: \text{ The amount of time when all } d_k \text{ in } S \text{ receive the pkt; } A \oplus B = \operatorname{span}(\mathbf{v} : \mathbf{v} \in A \cup B).$



The total # of to-be-sent symbols: [What r has heard] — [What r and d_1 both have heard]

The overheard info.:

[What *r* and d_2 have heard] – [What *r*, d_2 , and d_1 all have heard]





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- When K = 1, then rank $(\Omega_1) = \min(\mathcal{T}_1, N)$.
- When K = 2, we have $\operatorname{rank}(\Omega_1 \cap \Omega_2) = \operatorname{rank}(\Omega_1) + \operatorname{rank}(\Omega_2) - \operatorname{rank}(\Omega_1 \oplus \Omega_2)$ $= \min(\mathcal{T}_1, N) + \min(\mathcal{T}_2, N) - \min(\mathcal{T}_1 + \mathcal{T}_2 - \mathcal{T}_{\{1,2\}}, N)$



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• Our first thought was when K = 3, we should have $\operatorname{rank}\left(\bigcap_{k=1}^{3}\Omega_{k}\right) = \sum_{k=1}^{K}\operatorname{rank}\left(\Omega_{k}\right) - \sum_{i < j}\operatorname{rank}\left(\Omega_{i} \oplus \Omega_{j}\right) + \operatorname{rank}\left(\Omega_{1} \oplus \Omega_{2} \oplus \Omega_{3}\right)$



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An example is provided in the paper.


Main Result

• A *partition* of $\{1, \dots, K\}$ is a collection of disjoint subsets $\{S_m\} \stackrel{\Delta}{=} \{S_1, S_2, \dots, S_M\}$ such that $\bigcup_{m=1}^M S_m = \{1, \dots, K\}$.

Theorem 1 Define $(\cdot)^+ \stackrel{\Delta}{=} \max(\cdot, 0)$. For any receiving sets $\{\mathcal{R}_t : \forall t\}$, with sufficiently large $\mathsf{GF}(q)$ we have

$$\operatorname{rank}\left(\bigcap_{k=1}^{K}\Omega_{k}\right) = \max\left\{N - \sum_{m=1}^{M}\left(N - \mathcal{T}_{S_{m}}\right)^{+} : \forall \text{ partition } \{S_{m}\}\right\}$$



Proof of A Simplified Example

We prove a degenerate case in this presentation. Assume

$$N \ge \max_{k \in [K]} \mathcal{T}_k$$

 $\forall S \subseteq [K], \left(\sum_{k \in S} \mathcal{T}_k\right) - (|S| - 1)N \ge \mathcal{T}_S.$

• We will prove that rank $\left(\bigcap_{k=1}^{K}\Omega_{k}\right) = \sum_{k=1}^{K}\mathcal{T}_{k} - (K-1)N.$



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An illustrative example w. K = 3:

- N = 5 dimensional vector space.
- **•** 7 vectors with reception status 001, 010, 011, 100, 101, 110, 111.

•
$$T_1 = T_2 = T_3 = 4$$
, $T_{\{1,2\}} = T_{\{1,3\}} = T_{\{2,3\}} = 2$, and $T_{\{1,2,3\}} = 1$. Wang, ISIT 2011 - p. 10/1

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Wang, ISIT 2011 - p. 10

Rx 1 Rx 2

Rx 3







Wang, ISIT 2011 – p. 11/15



Wang, ISIT 2011 – p. 11/15







































 $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ depend on the other three vectors.





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Wang, ISIT 2011 – p. 11/15





[Koetter *et al.* 03, Ho *et al.* 06]: If we can find *deterministically* 7 vectors s.t. rank(100 110 101 111) 010 110 011 111) 010 011 010 011 010 011 111) 010 011 010 011 010 011 010 011 010 011 011 010 011 010 011 010 011 010 011

then for *almost all random choices* of the 7 vectors, $rank(\Omega_1 \oplus \Omega_2) = 5$.





Wang, ISIT 2011 – p. 12/15

[Koetter *et al.* 03, Ho *et al.* 06]: If we can find *deterministically* 7 vectors s.t. $\operatorname{Rx} 1$ $\operatorname{Rx} 2$ $\operatorname{Rx} 2$ Fully Spanned Cond. then for *almost all random choices* of the 7 vectors, $\operatorname{rank}(\Omega_1 \oplus \Omega_2) = 5$. The rank problem of RLNC is reduced to finding a deterministic assignment satisfying the Fully Spanned Cond.



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[New]: If we can find *deterministically* 10 vectors: \mathbf{v}_1 , \mathbf{v}_2 , \mathbf{v}_3 , plus 7, s.t. Rx 1 Rx 2 Fully Spanned rank(100 101 111 101 111 101 111 101 111



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It does not matter how we choose **v**₁, **v**₂, **v**₃, plus 7, and in what order we construct the vectors. Any deterministic construction will suffice!


































Intuition of the FS and FI Conds.

The Common Info. problem of RLNC is reduced to finding a deterministic assignment satisfying the Fully Spanned and Fully Intersected Conds.

- Then in RLNC, each message along an edge has the form of $M = (f_1(\mathbf{x}), f_2(\mathbf{x}), \cdots, f_N(\mathbf{x}))$

where $f_i(\mathbf{x})$ are polynomials of \mathbf{x} . [Koetter *et al.* 03].



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• When constructing the basis vectors of $\bigcap_k \Omega_k$, we need to solve linear equations (i.e., being in all marginal spaces), which results in the form $\mathbf{v} = \left(\frac{f_1(\mathbf{x})}{g_1(\mathbf{x})}, \frac{f_2(\mathbf{x})}{g_2(\mathbf{x})}, \cdots, \frac{f_N(\mathbf{x})}{g_N(\mathbf{x})}\right).$



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- [Fully Intersected] associates a deterministic \mathbf{x}_0 assignment with the corresponding fractional expressions; [Fully Spanned] guarantees both $g_n(\mathbf{x})$ and det ($[\mathbf{v}_i]$) are non-zero.

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Main Results
rank
$$\left(\bigcap_{k=1}^{K} \Omega_{k}\right) = \max\left\{N - \sum_{m=1}^{M} \left(N - \mathcal{T}_{S_{m}}\right)^{+} : \forall \text{ partition } \{S_{m}\}\right\}$$



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Main motivation:



The total # of to-be-sent symbols: rank $\left(\Omega_{r}^{[s_{1}]}\right)$ rank $\left(\Omega_{r}^{[s_{1}]} \cap \Omega_{d_{1}}^{[s_{1}]}\right)$ The overheard info.: rank $\left(\Omega_{r}^{[s_{1}]} \cap \Omega_{d_{2}}^{[s_{1}]}\right)$ rank $\left(\Omega_{r}^{[s_{1}]} \cap \Omega_{d_{2}}^{[s_{1}]} \cap \Omega_{d_{1}}^{[s_{1}]}\right)$



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Main motivation:

 $\begin{array}{c} X_1 \cdots X_{nR_1} & Y_1 \cdots Y_{nR_2} \\ \hline Random & Random \\ \hline PEC & PEC \\ \hline \hline \\ \hat{Y}_1 \cdots \hat{Y}_{nR_2} & \hat{X}_1 \cdots \hat{X}_{nR_1} \end{array} \begin{array}{c} \text{ra} \\ \hline \\ \hat{Y}_1 \cdots \hat{Y}_{nR_2} & \hat{X}_1 \cdots \hat{X}_{nR_1} \end{array}$

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■ Future works: (i) Arbitrary combinations: Ex: $(\Omega_1 \oplus \Omega_2) \cap \Omega_3$. (ii) Common Information of RLNC over multi-hop networks. Wang, ISIT 2011 - p. 15/15

