

# Coding and Limits for Wireless Distributed Computing

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- **Part I: Partially-Connected Networks**

- General interference alignment (IA) coding scheme
- A novel number-filling puzzle capturing the network structure
- The score of the puzzle solution reflects the sum rate

- **Part II: Wireless Distributed Computing (DC)**

- Focus on MapReduce-based wireless systems
- Exploiting structural parallels to apply IA scheme and enhance communication-computation tradeoff
- Information-theoretic converse bound proving near-optimality

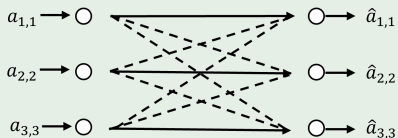
# Why Partially-Connected Networks?

- **Realistic Modeling:** Many practical systems (e.g., wireless, IoT, and fog networks) exhibit limited connectivity due to physical constraints, interference, or energy limits.
- **Broad Applications:** Distributed computing, D2D caching, and cooperative communication systems.
- **New Theoretical Questions:** Partial connectivity introduces new design and analysis challenges, e.g. topology-aware interference management, resource allocation.

# Partially Connected Channel

- Channel with  $K$  Tx's and  $K$  Rx's.
- $\mathbf{N} \in \{0, 1\}^{K \times K}$ : connectivity matrix.
- $\mathbf{M} \in \{0, 1\}^{K \times K}$ : message flow matrix.

## Example (Interference channel with $K = 3$ )

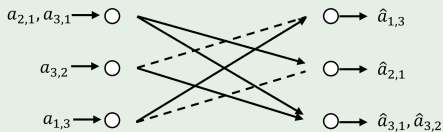


Corresponding matrices:

$$\mathbf{N} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}, \mathbf{M} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

# Partially Connected Channel

Example (A partially connected channel with  $K = 3$ )



Corresponding matrices:

$$\mathbf{N} = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}, \mathbf{M} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 0 \end{bmatrix}.$$

- Rx  $p$  observes

$$Y_p(t) = \sum_{\{q: \mathbf{N}[p,q]=1\}} H_{p,q}(t)X_q(t) + Z_p(t), \quad p \in [K], t \in [T]$$

- $T$  is the total blocklength.
- $\{H_{p,q}(t)\}$  are i.i.d. fading according to a bounded continuous distribution.
- Assume *perfect CSI*

- Tx  $q$  transmits a message  $a_{p,q}$  to Rx  $p$  for which  $\mathbf{M}[p, q] = 1$ .

$$\mathbf{X}_q \triangleq (X_q(1), \dots, X_q(T)) = f_q^{(T)}(\{a_{p,q} : \mathbf{M}[p, q] = 1\})$$

- Each Tx produces inputs with a power constraint  $P$ .
- Message  $a_{p,q}$  is uniformly distributed over  $[2^{\text{TR}_{p,q}}]$ .

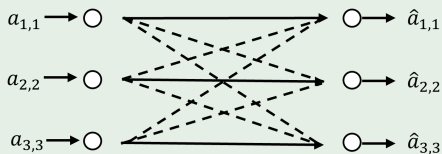
- *Sum Degrees of Freedom (SDoF):*

$$\text{SDoF} \triangleq \overline{\lim}_{P \rightarrow \infty} \sup_{\mathbf{R} \in \mathcal{C}(P)} \sum_{\mathbf{M}[p,q]=1} \frac{R_{p,q}}{\log P}.$$

*Capacity region*  $\mathcal{C}(P)$  is set of tuple  $(R_{p,q})$  s.t.  $p(\text{error}) \rightarrow 0$  for given power  $P$ .

# Interference Alignment (IA) scheme for Interference Channels

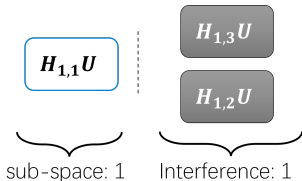
Example (An interference channel with  $K = 3$ )



- Encode each message  $a_{q,q}$  to  $\eta^\Gamma$ -length codeword  $\mathbf{b}_{q,q}$ , and pre-multiply  $\mathbf{U}$ :

$$\mathbf{X}_q = \mathbf{U}\mathbf{b}_{q,q}$$

- By designing  $\mathbf{U}$ , Rx1 subspace (Rx2, Rx3 are similar):



# Interference Alignment (IA) scheme for interference channel

- We choose  $\mathcal{H}$  as the set of all channel coefficients of interference links:

$$\mathbf{H}_{1,2}, \mathbf{H}_{1,3}, \mathbf{H}_{2,1}, \mathbf{H}_{2,3}, \mathbf{H}_{3,1}, \mathbf{H}_{3,2}.$$

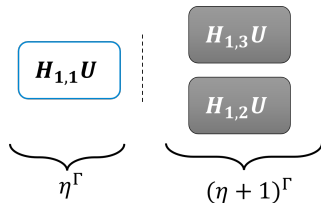
- The matrix  $\mathbf{U}$  is

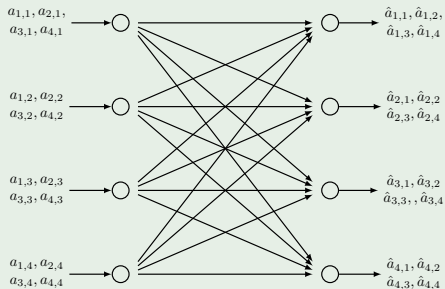
$$\begin{aligned} \mathbf{U} &= \left[ \mathbf{1}, \mathbf{H}_{1,2}\mathbf{1}, \mathbf{H}_{1,2}^2\mathbf{1}, \dots, \mathbf{H}_{1,2}^{\eta}\mathbf{1}, \mathbf{H}_{1,2}\mathbf{H}_{1,3}\mathbf{1}, \dots, \mathbf{H}_{1,2}^{\eta}\mathbf{H}_{1,3}^{\eta}\mathbf{1}, \dots \right] \\ &= \left[ \prod_{\mathbf{H}_{p,q} \in \mathcal{H}} \mathbf{H}_{p,q}^{\alpha_{p,q}} \cdot \mathbf{1} : \forall \alpha \in [\eta]^{\Gamma} \right], \quad \text{with } \Gamma = |\mathcal{H}|. \end{aligned}$$

- For all  $\mathbf{H}_{p,q} \in \mathcal{H}$ ,

$$\text{span}(\mathbf{H}\mathbf{U}) \subseteq \text{span}(\mathbf{W}), \quad \mathbf{W} = \left[ \prod_{\mathbf{H}_{p,q} \in \mathcal{H}} \mathbf{H}_{p,q}^{\alpha_{p,q}} \cdot \mathbf{1} : \forall \alpha \in [\eta + 1]^{\Gamma} \right].$$

- For all  $\mathbf{H}_{p,p} \notin \mathcal{H}$ , e.g.  $\mathbf{H}_{1,1}$ ,  $\mathbf{H}\mathbf{U}$  has full column rank,  $\text{span}(\mathbf{H}\mathbf{U})$  and  $\text{span}(\mathbf{W})$  are **linearly independent**.



Example (X-channel with  $K = 4$ )

Corresponding matrices:

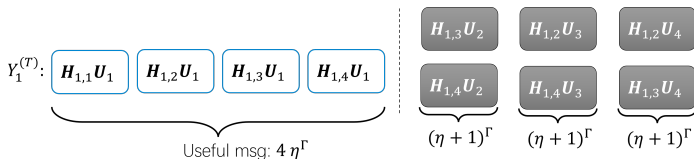
$$\mathbf{M} = \mathbf{N} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}.$$

- Encode  $a_{p,q}$  to  $\eta^\Gamma$ -length codeword  $\mathbf{b}_{p,q}$ .

- $\mathbf{U}_p = \left[ \prod_{\mathbf{H} \in \mathcal{H}_p} \mathbf{H}^{\alpha_{\mathbf{H}}} \cdot \Xi_p : \forall \alpha \in [\eta]^\Gamma \right]$  multiply  $\{\mathbf{b}_{p,1}, \dots, \mathbf{b}_{p,4}\}$ .

$$\begin{aligned} X_1^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,1} + \mathbf{U}_2 \mathbf{b}_{2,1} + \mathbf{U}_3 \mathbf{b}_{3,1} + \mathbf{U}_4 \mathbf{b}_{4,1}, \\ X_2^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,2} + \mathbf{U}_2 \mathbf{b}_{2,2} + \mathbf{U}_3 \mathbf{b}_{3,2} + \mathbf{U}_4 \mathbf{b}_{4,2}, \\ X_3^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,3} + \mathbf{U}_2 \mathbf{b}_{2,3} + \mathbf{U}_3 \mathbf{b}_{3,3} + \mathbf{U}_4 \mathbf{b}_{4,3}, \\ X_4^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,4} + \mathbf{U}_2 \mathbf{b}_{2,4} + \mathbf{U}_3 \mathbf{b}_{3,4} + \mathbf{U}_4 \mathbf{b}_{4,4}. \end{aligned}$$

- By choose matrices  $\mathbf{U}_p$  such that <sup>1</sup>: (Rx 1, 2, 3, 4 are similar)



<sup>1</sup>V. R. Cadambe and S. A. Jafar, "Interference Alignment and the Degrees of Freedom of Wireless X Networks," IEEE Trans. Inform. Theory, Sep. 2009.

# Matrix representation of IA schemes

- A matrix  $\mathbf{G}$  is applied to describe the precoding matrix allocation.
  - $\mathbf{G}[p, q] = \mathbf{G}[p', q'] = g > 0$ :  $\mathbf{b}_{p,q}$  and  $\mathbf{b}_{p',q'}$  are premultiplied by the same  $\mathbf{U}_g$ .
  - $\mathbf{G}[p, q] = 0$ :  $\mathbf{b}_{p,q}$  is not transmitted.

$$\begin{aligned}X_1^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,1} + \mathbf{U}_2 \mathbf{b}_{2,1} + \mathbf{U}_3 \mathbf{b}_{3,1} + \mathbf{U}_4 \mathbf{b}_{4,1}, \\X_2^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,2} + \mathbf{U}_2 \mathbf{b}_{2,2} + \mathbf{U}_3 \mathbf{b}_{3,2} + \mathbf{U}_4 \mathbf{b}_{4,2}, \\X_3^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,3} + \mathbf{U}_2 \mathbf{b}_{2,3} + \mathbf{U}_3 \mathbf{b}_{3,3} + \mathbf{U}_4 \mathbf{b}_{4,3}, \\X_4^{(T)} &= \mathbf{U}_1 \mathbf{b}_{1,4} + \mathbf{U}_2 \mathbf{b}_{2,4} + \mathbf{U}_3 \mathbf{b}_{3,4} + \mathbf{U}_4 \mathbf{b}_{4,4}.\end{aligned}$$



$$\mathbf{G} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \end{bmatrix}$$

## Calculate SDoF with Matrix representation

- $\|\mathbf{G}[:, p]\|_0$ : the number of non-zero entries in the  $p$ -th column (codewords for Rx  $p$ ) of  $\mathbf{G}$ .
- $g^{(p)}$ : the number of different non-zero integers in the interfering codeword submatrix, which is

$$\mathbf{G}^{(p)} \triangleq \mathbf{G}[p' \neq p, \{q: \mathbf{N}[p, q] = 1\}].$$

- DoF of Rx  $p$  is given by:

$$\text{DoF}_p = \frac{\|\mathbf{G}[p, :]\|_0}{\|\mathbf{G}[p, :]\|_0 + g^{(p)}},$$

For  $p = 1$  ( $p \in \{2, 3, 4\}$  are similar),

$$\mathbf{G}[:, 1] = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \Rightarrow \|\mathbf{G}[:, 1]\|_0 = 4, \quad \mathbf{G}^{(1)} = \begin{bmatrix} 2 & 3 & 4 \\ 2 & 3 & 4 \\ 2 & 3 & 4 \\ 2 & 3 & 4 \end{bmatrix} \Rightarrow g^{(1)} = 3.$$

$$\text{SDoF} = \frac{16}{4+3} = 16/7.$$

# Main Result: Correspondence between $\mathbf{G}$ s and IA schemes

$\mathbf{G}$  should satisfy the following two requirements:

- $\mathbf{G}$  inherits zeros from  $\mathbf{M}$ . I.e.,  $\mathbf{M}[p, q] = 0 \Rightarrow \mathbf{G}[p, q] = 0$ .
- All non-zero entries in a column are different. I.e.,  $\mathbf{G}[p, q] = \mathbf{G}[p', q] \Rightarrow \mathbf{G}[p, q] = 0$ .

## Theorem

For an  $(\mathbf{M}, \mathbf{N})$ -channel and a matrix  $\mathbf{G}$  satisfying the above two points, Sum-DoF of the network is lower bounded by:

$$\text{Sum-DoF} \geq \frac{\|\mathbf{G}\|_0}{\max_{p \in [K]} \{ \|\mathbf{G}[p, :]\|_0 + g^{(p)} \}},$$

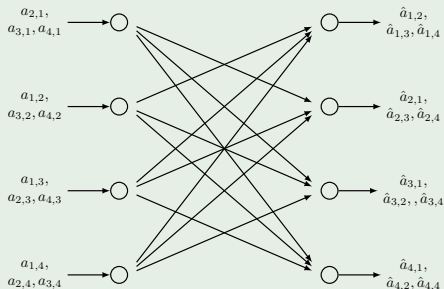
where

- $\|\mathbf{G}[p, :]\|_0$ : the number of non-zero entries in the  $p$ -th row of  $\mathbf{G}$ .
- $g^{(p)}$ : the number of different non-zeros integers in the submatrix

$$\mathbf{G}^{(p)} \triangleq \mathbf{G}[p' \neq p, \{q: \mathbf{N}[p, q] = 1\}].$$

# Partially Connected Channel

## Example (Partially-connected X-channel)



Corresponding matrices:

$$\mathbf{M} = \mathbf{N} = \begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{bmatrix}.$$

## Example (Classic IA)

$$\mathbf{M} = \mathbf{N} = \begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{bmatrix}, \mathbf{G} = \begin{bmatrix} 0 & 2 & 3 & 4 \\ 1 & 0 & 3 & 4 \\ 1 & 2 & 0 & 4 \\ 1 & 2 & 3 & 0 \end{bmatrix},$$

For  $p = 1$  ( $p \in \{2, 3, 4\}$  are similar),

$$\|\mathbf{G}[:, 1]\|_0 = 3, \quad \mathbf{G}^{(1)} = \begin{bmatrix} 0 & 3 & 4 \\ 2 & 0 & 4 \\ 2 & 3 & 0 \end{bmatrix} \Rightarrow g^{(1)} = 3.$$

$$\text{Sum-DoF} = \frac{12}{3+3} = 2.$$

## Proposed Scheme for Partially-Connected Channel with $K = 4$

- To ensure decodability, all non-zero entries in a row should be **different**.

### Example (IA with new allocation scheme)

- Encode all message **except**  $(p, q) = (3, 4)$ .
- Only apply 3 precoding matrices.

$$\mathbf{M} = \mathbf{N} = \begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{bmatrix}, \mathbf{G} = \begin{bmatrix} 0 & 2 & 3 & 1 \\ 1 & 0 & 3 & 2 \\ 1 & 2 & 0 & 0 \\ 1 & 2 & 3 & 0 \end{bmatrix}$$

For  $p = 1$  ( $p \in \{2, 3\}$  are similar),

$$\|\mathbf{G}[:, 1]\|_0 = 3, \quad \mathbf{G}^{(1)} = \begin{bmatrix} 0 & 3 & 2 \\ 2 & 0 & 0 \\ 2 & 3 & 0 \end{bmatrix} \Rightarrow g^{(1)} = 2.$$

For  $p = 4$ ,

$$\|\mathbf{G}[:, 4]\|_0 = 2, \quad \mathbf{G}^{(4)} = \begin{bmatrix} 0 & 2 & 3 \\ 1 & 0 & 3 \\ 1 & 2 & 0 \end{bmatrix} \Rightarrow g^{(4)} = 3.$$

$$\text{Sum-DoF} = \frac{11}{2+3} = 11/5.$$

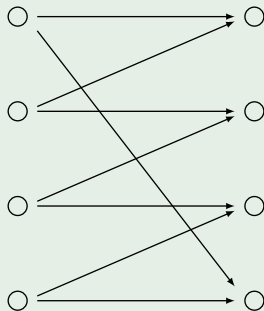
## General Results for Symmetric Regular Channels

- Consider symmetric channels with  $K$  user, where each Tx is connected to  $m = n \in [K]$  Rx(s), and each Tx sends an independent message to each connected Rx:

$$\mathbf{N}[p, q] = \begin{cases} 1 & \text{if } (p - q) \bmod K < n \\ 0 & \text{otherwise} \end{cases}, \quad \mathbf{M}[p, q] = \begin{cases} 1 & \text{if } (p - q) \bmod K < m \\ 0 & \text{otherwise} \end{cases}.$$

Example (An example with  $K = 4$ ,  $m = n = 2$ )

$$\mathbf{M} = \mathbf{N} = \begin{bmatrix} 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix},$$



# Numerical Results for Symmetric Regular Channels

## Theorem (Bounds for regular channels)

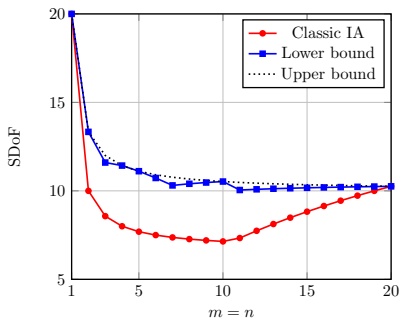
For the channel with matrices  $\mathbf{M}, \mathbf{N}$   $m < n$ , Sum-DoF =  $\frac{K}{2}$ .

For  $m = n$ , we have the following lower bound:

$$\text{Sum-DoF}_{\text{Lb}} \geq \frac{Km - (K \bmod m) + \epsilon}{2m - 1}, \text{ where } \epsilon = \begin{cases} 1, & m = (K + 1)/2 \text{ and } K \text{ is odd,} \\ 0, & \text{otherwise.} \end{cases}$$

Upper bound:

$$\text{Sum-DoF}_{\text{Lb}} = \frac{K \cdot m}{2m - 1}.$$



- Introduced a novel number-filling puzzle as an intuitive design tool for IA scheme.
- Proposed a improved IA scheme for partially-connected networks.

### **Outlook:**

- Extend the "puzzle" to networks with cooperation

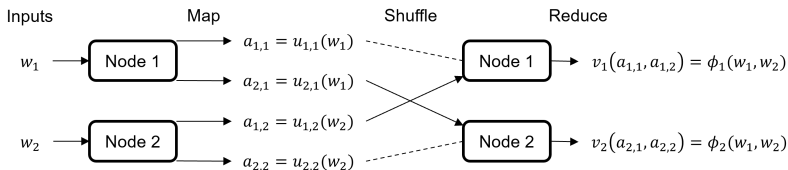
- Distributed system is a group of networked computers. These computers communicate and coordinate their actions to achieve a common goal.
  
- As components in distributed systems are physically separated and wireless smart devices have become increasingly dominant, studying wireless coding schemes for distributed systems is crucial.
  - Improved wireless coding schemes can significantly reduce the overall execution time of distributed computing systems.
  - Maintaining accuracy is critical for distributed estimation systems.
  - Efficient data synchronization is essential for distributed storage systems to ensure consistency and reliability.

# Wireless MapReduce Framework

- $\phi_k(w_1, \dots, w_N) = v_k(\underbrace{u_{k,1}(w_1)}_{a_{k,1}}, \dots, \underbrace{u_{k,N}(w_N)}_{a_{k,N}})$ ,  $k \in [Q]$
- **Map phase:**  $\mathcal{M}_k \subseteq [N]$  is assigned to node  $k$ , it computes intermediate values (IVA)  $a_{k,n}$ .
- **Shuffle phase:** Exchange IVAs over wireless channel. Assume nodes are **full-duplex**.

$$Y_k(t) = \sum_{\ell \in [K] \setminus k} H_{k,\ell}(t) X_\ell(t) + Z_k(t), \quad t \in [T],$$

- **Reduce phase:** Apply reduce functions.



# Wireless DC System Model: Performance Measures

The performance is measured by two factors:

- The *computation load*  $r \in [1, K]$

$$r \triangleq \sum_{p \in [K]} \frac{|\mathcal{M}_p|}{N},$$

- The *normalized delivery time (NDT)*  $\Delta \in [0, 1]$

$$\Delta \triangleq \lim_{P \rightarrow \infty} \lim_{A \rightarrow \infty} \frac{T}{Q \cdot N \cdot A / \log P}.$$

Objective: optimal computation-NDT tradeoff  $\Delta^*(r)$

$$\Delta^*(r) \triangleq \min_{\{f_p^{(T)}, g_{q,i}^{(T)}\}, \{\mathcal{M}_p\}} \Delta$$

Sum Degrees of Freedom (SDoF):

$$\text{SDoF} \triangleq \overline{\lim}_{P \rightarrow \infty} \sup_{\mathbf{R} \in \mathcal{C}(P)} \sum_{k \in [\tilde{K}]} \sum_{p \in [K] \setminus \mathcal{T}_k} \frac{R_{p,k}(P)}{\log P}.$$

Capacity region  $\mathcal{C}(P)$  is set of tuple  $(R_{p,k}(P))$  s.t.  $p(\text{error}) \rightarrow 0$ .

- Apply our IA scheme for the channel.

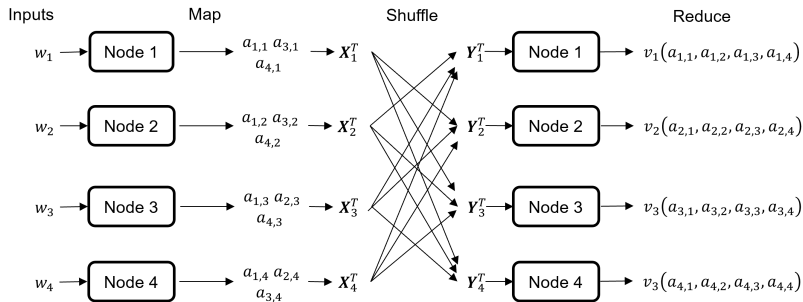
## Lemma

$Q$  is a multiple of  $K$ . As  $N \rightarrow \infty$ ,

$$\Delta^*(r) \leq \text{lowc} \left( (K, 0) \cup \left\{ \left( r, \frac{1-r/K}{\text{SDoF}_{\text{Lb}}} \right) : 1 \leq r < K \text{ and } r|K \right\} \right),$$

where  $\text{lowc}(\cdot)$  denotes the lower-convex envelope.

# IA for Wireless DC System Model with $r = 1$



- **File assignment & Map phase:** Each node is assigned an output function and an input file.
- **Shuffle phase:** Channel is equivalent to a partially connected channel with

$$\mathbf{M} = \mathbf{N} = \begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{bmatrix}.$$

## $r > 1$ : Combinatorial File Assignment Scheme

- Assign a distinct input file to each group of  $r$  nodes.<sup>2</sup>
- Redesigned the precoding matrix allocation scheme.

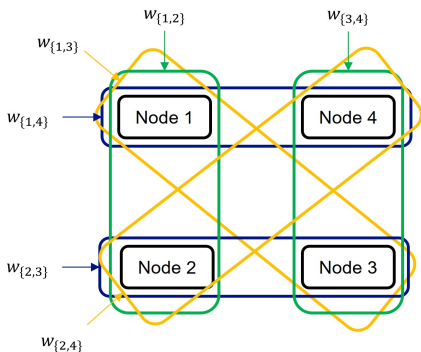


Figure: An example with  $K = 4$ ,  $r = 2$

<sup>2</sup>S. Li et al., "A Fundamental Tradeoff Between Computation and Communication in Distributed Computing," IEEE Trans. Inform. Theory, Jan. 2018.

## Proposed Scheme for $K = 4, r = 2$ : Encoding

- Send all messages  $M_{k,\mathcal{T}}^i$  except all message from node 4 to node 1 i.e.  $j = 1, k = 4$ .
  - To Node 1 we send

$$M_{2,\{2,3\}}^1, M_{3,\{2,3\}}^1, M_{2,\{2,4\}}^1, M_{3,\{3,4\}}^1;$$

- To Node 2 we send

$$M_{1,\{1,3\}}^2, M_{3,\{1,3\}}^2, M_{1,\{1,4\}}^2, M_{4,\{1,4\}}^2, M_{3,\{3,4\}}^2, M_{4,\{3,4\}}^2;$$

- To Node 3 we send

$$M_{1,\{1,2\}}^3, M_{2,\{1,2\}}^3, M_{1,\{1,4\}}^3, M_{4,\{1,4\}}^3, M_{2,\{2,4\}}^3, M_{4,\{2,4\}}^3;$$

- To Node 4 we send

$$M_{1,\{1,2\}}^4, M_{2,\{1,2\}}^4, M_{1,\{1,3\}}^4, M_{3,\{1,3\}}^4, M_{2,\{2,3\}}^4, M_{3,\{2,3\}}^4.$$

## Proposed Scheme for $K = 4, r = 2$ : Encoding

- Encode the message  $M_{k,\mathcal{T}}^j$  to  $\eta^\Gamma$ -length Gaussian codeword  $\mathbf{b}_{k,\mathcal{T}}^j$ .
- $\mathbf{b}_{k,\mathcal{T}}^j$  is precoded by
  - if  $j = 1, \mathbf{U}_{\mathcal{T}}$
  - if  $1 \in \mathcal{T}, \mathbf{U}_{\mathcal{T} \setminus \{1\} \cup \{j\}}$
  - if  $1 \notin \mathcal{T}$  and  $j \neq 1, \mathbf{U}_{\mathcal{T} \setminus \{k\} \cup \{j\}}$

$$\begin{aligned} \mathbf{X}_1 &= \mathbf{U}_{\{2,3\}} \left( \mathbf{b}_{1,\{1,3\}}^2 + \mathbf{b}_{1,\{1,2\}}^3 \right) && + \mathbf{U}_{\{2,4\}} \left( \mathbf{b}_{1,\{1,4\}}^2 + \mathbf{b}_{1,\{1,2\}}^4 \right) \\ &+ \mathbf{U}_{\{3,4\}} \left( \mathbf{b}_{1,\{1,4\}}^3 + \mathbf{b}_{1,\{1,3\}}^4 \right), \\ \mathbf{X}_2 &= \mathbf{U}_{\{2,3\}} \left( \mathbf{b}_{2,\{2,3\}}^1 + \mathbf{b}_{2,\{1,2\}}^3 \right) && + \mathbf{U}_{\{2,4\}} \left( \mathbf{b}_{2,\{2,4\}}^1 + \mathbf{b}_{2,\{1,2\}}^4 \right) \\ &+ \mathbf{U}_{\{3,4\}} \left( \mathbf{b}_{2,\{2,4\}}^3 + \mathbf{b}_{2,\{2,3\}}^4 \right), \end{aligned}$$

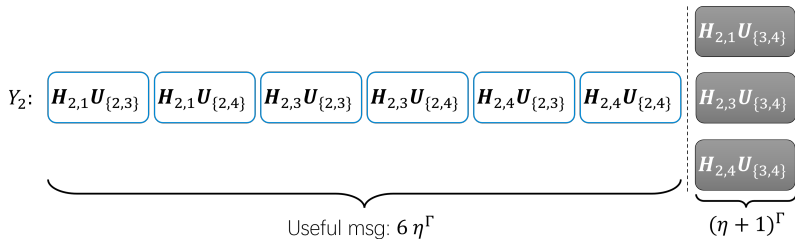
## Proposed Scheme for $K = 4, r = 2$ : Encoding

- Encode the message  $a_{p,k}$  to  $\eta^\Gamma$ -length Gaussian codeword  $\mathbf{b}_{p,k}$  except  $(p, k) = (1, 4)$ .
- $\mathbf{b}_{k,\mathcal{T}}^j$  is precoded by
  - if  $j = 1, \mathbf{U}_{\mathcal{T}}$
  - if  $1 \in \mathcal{T}, \mathbf{U}_{\mathcal{T} \setminus \{1\} \cup \{j\}}$
  - if  $1 \notin \mathcal{T}$  and  $j \neq 1, \mathbf{U}_{\mathcal{T} \setminus \{k\} \cup \{j\}}$

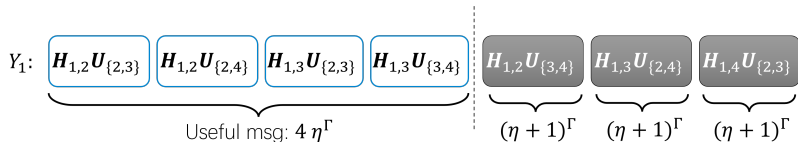
$$\begin{aligned} \mathbf{X}_3 &= \mathbf{U}_{\{2,3\}} \left( \mathbf{b}_{3,\{2,3\}}^1 + \mathbf{b}_{2,\{1,3\}}^2 \right) && + \mathbf{U}_{\{2,4\}} \left( \mathbf{b}_{3,\{3,4\}}^2 + \mathbf{b}_{3,\{2,3\}}^4 \right) \\ &+ \mathbf{U}_{\{3,4\}} \left( \mathbf{b}_{3,\{3,4\}}^1 + \mathbf{b}_{3,\{1,3\}}^4 \right), \\ \mathbf{X}_4 &= \mathbf{U}_{\{2,3\}} \left( \mathbf{b}_{4,\{3,4\}}^2 + \mathbf{b}_{4,\{2,4\}}^3 \right) && + \mathbf{U}_{\{2,4\}} \mathbf{b}_{4,\{1,4\}}^2 \\ &+ \mathbf{U}_{\{3,4\}} \mathbf{b}_{4,\{1,4\}}^3 \end{aligned}$$

## Proposed Scheme for $K = 4, r = 2$ : Decoding

- Choose matrices  $\mathbf{U}_{\{2,3\}}, \mathbf{U}_{\{2,4\}}, \mathbf{U}_{\{3,4\}}$  such that [3]: (Rx 2,3,4 are similar)



- At Rx 1



- Choose  $T = 4\eta^\Gamma + 3(\eta + 1)^\Gamma$

$$\text{SDoF} = \lim_{\eta \rightarrow \infty} \frac{3 \times 6\eta^\Gamma + 4\eta^\Gamma}{4\eta^\Gamma + 3(\eta + 1)^\Gamma} \approx \frac{3 \times 6}{7} + \frac{4}{7} = \frac{22}{7}$$

# Upper Bound for Wireless DC System

## Theorem

The computation-NDT tradeoff  $\Delta^*(r)$  is upper-bounded as:

$$\Delta^*(r) \leq \text{lowc}(\Delta_{\text{Ub}}(r)).$$

with

$$\Delta_{\text{Ub}}(r) \triangleq \begin{cases} \min_{i \in \{1,2\}} \Delta_{\text{Ub},i}(r) & \text{if } r < K/2 \\ \frac{1}{K} \left(1 - \frac{r}{K}\right) & \text{if } r \geq K/2 \end{cases}.$$

For fixed  $K$ , define for each  $r \in [\lceil K/2 \rceil - 1]$ :

$$\Delta_{\text{Ub},1}(r) \triangleq \left(1 - \frac{r}{K}\right) \cdot \frac{r(K-1) + K - r - 1}{r(K-1)^2 + r(K-2)}.$$

For  $K = 5$  and  $r = 2$ , define

$$\Delta_{\text{Ub},2}(r) \triangleq \left(1 - \frac{r}{K}\right) \cdot \frac{7}{30}$$

and for all odd values  $K \geq 7$  and  $r = (K-1)/2$ , set:

$$\Delta_{\text{Ub},2}(r) \triangleq \frac{1}{K} \left(1 - \frac{r}{K}\right) \left(1 + \frac{1}{(K-r-1)(K-1)}\right).$$

For all other values of  $r$  and  $K$ , set  $\Delta_{\text{Ub},2}(r) = \infty$ .

# Proof of Lower Bound

## Lemma

Consider two disjoint sets  $\mathcal{T}$  and  $\mathcal{R}$  of same size, and  $\mathcal{F} \triangleq [K] \setminus (\mathcal{R} \cup \mathcal{T})$ . Let  $\mathcal{M} \subseteq [N]$  be the set of files known only to nodes  $\mathcal{T}$  but not to any other node and partition the set of all IVAs  $\mathcal{A}$  into the following disjoint subsets:

$$\mathcal{W}_r \triangleq \{a_{j,m} \mid j \in \mathcal{R}, m \in [N] \setminus \mathcal{M}_j\}, \quad \mathcal{W}_t \triangleq \{a_{j,m} \mid j \in \mathcal{T}, m \in \mathcal{M} \setminus \mathcal{M}_j\}.$$

For any sequence of distributed computing systems:

$$d \triangleq \overline{\lim}_{P \rightarrow \infty} \overline{\lim}_{T \rightarrow \infty} \frac{A}{T \log P} \leq \frac{|\mathcal{T}|}{|\mathcal{W}_t| + |\mathcal{W}_r|}$$

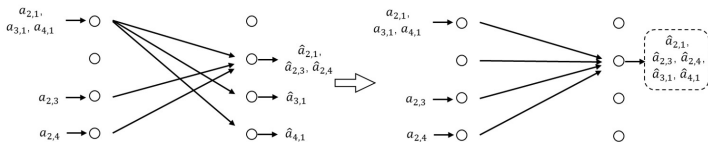


Figure: Example of Lemma with  $K = 4$  and  $r = 1$ .

- The lower bound can be obtained by linear combination of the Lemma with different  $(\mathcal{T}, \mathcal{R})$

## Theorem

The computation-NDT tradeoff  $\Delta^*(r)$  is lower-bounded as:

$$\Delta_{\text{Lb}}(r) \leq \Delta^*(r).$$

$$\Delta_{\text{Lb}}(r) \triangleq \begin{cases} \frac{1}{K} \left( 2 - \frac{3}{K} \right) & \text{if } r = 1, \\ \frac{1}{K} \left( 1 - \frac{r}{K} + \max_{t \in \llbracket K/2 \rrbracket} \text{lowc}(C_t(r)) \right) & \text{if } r \in \left( 1, \lceil \frac{K}{2} \rceil \right), \\ \frac{1}{K} \left( 1 - \frac{r}{K} \right) & \text{if } r \in \left[ \lceil \frac{K}{2} \rceil, K \right], \end{cases}$$

where for any  $t \in \llbracket K/2 \rrbracket$ :

$$C_t(i) = \frac{\binom{K-i}{t-i}}{\binom{K}{t} \cdot t} \cdot (K - 2t),$$

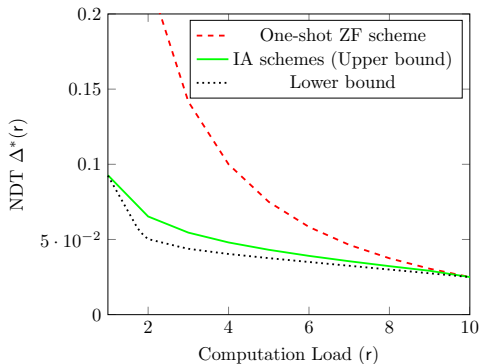


Figure: Lower bound and upper bounds on  $\Delta^*(r)$  when  $K = 20$ .

# Wireless DC System with Half-duplex Nodes

- We compare the bounds for wireless DC system with full-duplex nodes and half-duplex nodes.

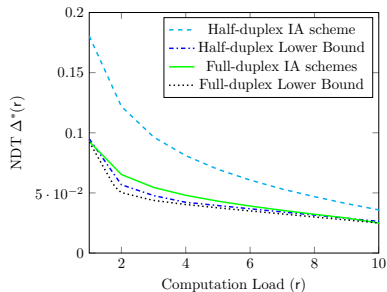
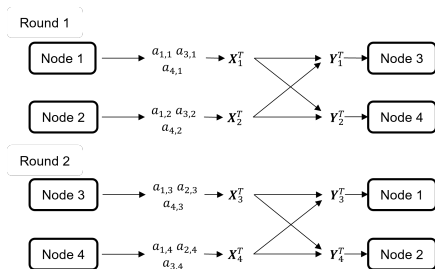


Figure: Bounds on  $\Delta^*(r)$  when  $K = 20$ .

- Revealed parallels between wireless DC and partially connected network.
- Achieved near-optimal communication-computation tradeoff.
- Developed information-theoretic converse bound for both full-duplex and half-duplex systems.

### **Outlook:**

- Analyze how input file assignment affects system performance.
- Apply to specific functions: binary, linear separable, etc.
- Practical implementation: finite block length, imperfect CSI, etc.

THANK YOU!

### Journal papers:

- **Y. Bi**, M. Wigger, Y. Wu, "Normalized Delivery Time of Wireless MapReduce", in *IEEE Transaction on Information Theory*, vol. 70, no. 10, pp. 7005-7022, Oct. 2024, doi: 10.1109/TIT.2024.3423710.
- Z. Huang, K. Yuan, S. Ma, **Y. Bi** and Y. Wu, "Coded Computing for Half-Duplex Wireless Distributed Computing Systems via Interference Alignment," in *IEEE Transactions on Wireless Communications*, vol. 23, no. 11, pp. 17399-17414, Nov. 2024, doi: 10.1109/TWC.2024.3453403.

### Conferences:

- **Y. Bi**, P. Ciblat, M. Wigger, and Y. Wu, "DoF of a Cooperative X-Channel with an Application to Distributed Computing," in *2022 IEEE International Symposium on Information Theory (ISIT)*, Jun. 2022.
- **Y. Bi**, M. Wigger, Y. Wu, "A New Interference-Alignment Scheme for Wireless MapReduce", in *2023 IEEE Globecom*, Dec. 2023.
- **Y. Bi**, Y. Wu, C. Hua, "DoF Analysis for (M, N)-Channels through a Number-Filling Puzzle", in *2024 IEEE ISIT*, Jul. 2024.

- $\mathcal{H}$ : Set of all channel coefficients
- $\mathcal{W}_t, \mathcal{W}_r$ : Target and recoverable IVAs
- $\mathcal{W}_c \triangleq \mathcal{A} \setminus (\mathcal{W}_t \cup \mathcal{W}_r)$
- $\mathbf{Y}_{\mathcal{A}} \triangleq [\mathbf{Y}_j]_{j \in \mathcal{A}}$

- Entropy Decomposition

$$\begin{aligned} H(\mathcal{W}_t, \mathcal{W}_r) &= H(\mathcal{W}_t, \mathcal{W}_r | \mathcal{W}_c, \mathcal{H}) \\ &= I(\mathcal{W}_t, \mathcal{W}_r; \mathbf{Y}_{\mathcal{R}} | \mathcal{W}_c, \mathcal{H}) + H(\mathcal{W}_t, \mathcal{W}_r | \mathbf{Y}_{\mathcal{R}}, \mathcal{W}_c, \mathcal{H}) \\ &\leq h(\mathbf{Y}_{\mathcal{R}} | \mathcal{W}_c, \mathcal{H}) - h(\mathbf{Z}_{\mathcal{R}}) + T\epsilon_T + H(\mathcal{W}_t | \mathcal{W}_r, \mathcal{W}_c, \mathbf{Y}_{\mathcal{R}}, \mathcal{H}) \end{aligned}$$

# Bounding Remaining Entropy

- By Fano's inequality

$$\begin{aligned} H(\mathcal{W}_t | \mathcal{W}_r, \mathcal{W}_c, \mathbf{Y}_{\mathcal{R}}, \mathcal{H}) &\leq I(\mathcal{W}_t; \mathbf{Y}_{\mathcal{F}} | \mathcal{W}_r, \mathcal{W}_c, \mathbf{Y}_{\mathcal{R}}, \mathcal{H}) + T\epsilon'_T \\ &= h(\bar{\mathbf{Y}}_{\mathcal{F}} | \bar{\mathbf{Y}}_{\mathcal{R}}, \mathcal{H}) - h(\mathbf{Z}_{\mathcal{F}}) + T\epsilon'_T \end{aligned}$$

- $\bar{\mathbf{Y}}_j \triangleq \mathbf{H}_{j, \mathcal{T}} \mathbf{X}_{\mathcal{T}} + \mathbf{Z}_j$
- If  $\mathbf{H}_{\mathcal{R}, \mathcal{T}}$  is invertible:

- Cleaned Signals and Residual Noise

$$\begin{aligned} h(\bar{\mathbf{Y}}_{\mathcal{F}} | \bar{\mathbf{Y}}_{\mathcal{R}}, \mathcal{H}) &\leq h(\bar{\mathbf{Z}}_{\mathcal{F}}) + \mathbb{P}(E = 0)h(\bar{\mathbf{Y}}_{\mathcal{F}} | \bar{\mathbf{Y}}_{\mathcal{R}}, \mathcal{H}, E = 0) \\ &= h(\bar{\mathbf{Z}}_{\mathcal{F}}) \text{ since } \mathbb{P}(E = 0) = 0 \end{aligned}$$

- Final Bound and Conclusion

$$\begin{aligned} H(\mathcal{W}_t, \mathcal{W}_r) &\leq h(\mathbf{Y}_{\mathcal{R}} | \mathcal{H}) - h(\mathbf{Z}_{\mathcal{R}}) + h(\bar{\mathbf{Z}}_{\mathcal{F}}) - h(\mathbf{Z}_{\mathcal{F}}) + T(\epsilon_T + \epsilon'_T) \\ &\leq T|\mathcal{R}| \log(P) + TC_{T, \mathcal{H}} \\ \Rightarrow \frac{|\mathcal{W}_t| + |\mathcal{W}_r|}{T \log P} &\leq |\mathcal{R}| = |\mathcal{T}|, \text{ as } P \rightarrow \infty \end{aligned}$$