

**On the Optimality of Beamforming for
Multi-User MISO
Interference Channels with Single-User Detection**

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Multi-user MISO Interference Channel (IC) Model

$$Y_i = \sum_{j=1, j \neq i}^m \mathbf{h}_{ji}^T \mathbf{x}_j + N_i$$

- m transmitter-receiver pairs.
- \mathbf{x}_i : $t_i \times 1$ transmitted signal at transmitter i . $\|\mathbf{x}_i\|^2 \leq P_i$.
- Y_i : received scalar signal at receiver i .
- $N_i \sim \mathcal{N}(0, 1)$.
- \mathbf{h}_{ji} : $t_j \times 1$ channel vector from transmitter j to receiver i .
- Transmitter i knows all channel states, receiver i only knows \mathbf{h}_{ii} .

Single-User Detection Rate Region

$$\begin{aligned}
 & \max \sum_{i=1}^m \mu_i R_i && (\mu_i \geq 0) \\
 & \text{subject to} && R_i = \frac{1}{2} \log \left(1 + \frac{\mathbf{h}_{ii}^T \mathbf{S}_i \mathbf{h}_{ii}}{1 + \sum_{j=1, j \neq i}^m \mathbf{h}_{ji}^T \mathbf{S}_j \mathbf{h}_{ji}} \right) \\
 & && \text{tr}(\mathbf{S}_i) \leq P_i, \quad \mathbf{S}_i \succeq \mathbf{0}, \quad i = 1, \dots, m.
 \end{aligned}$$

- $\mathbf{x}_i \sim \mathcal{N}(\mathbf{0}, \mathbf{S}_i)$.
- Receiver i decodes \mathbf{x}_i from Y_i by treating interference as noise.
- Single-user detection (SUD) rate region is defined as

$$\bigcup_{\text{all } \boldsymbol{\mu} = [\mu_1, \dots, \mu_m]} \{R_1, \dots, R_m\}$$

Problem Reformulation

- Why: Non-convex optimization problem.

$$\mu_1 \log \left(1 + \frac{\mathbf{h}_{11}^T \mathbf{S}_1 \mathbf{h}_{11}}{1 + \mathbf{h}_{21}^T \mathbf{S}_2 \mathbf{h}_{21}} \right) + \mu_2 \log \left(1 + \frac{\mathbf{h}_{22}^T \mathbf{S}_2 \mathbf{h}_{22}}{1 + \mathbf{h}_{12}^T \mathbf{S}_1 \mathbf{h}_{12}} \right)$$

- How: non-convex problem \Rightarrow convex problem

$$\begin{aligned} & \text{maximize} && \mu_1 \log \left(1 + \frac{\mathbf{h}_{11}^T \mathbf{S}_1 \mathbf{h}_{11}}{1 + z_2^2} \right) + \mu_2 \log \left(1 + \frac{\mathbf{h}_{22}^T \mathbf{S}_2 \mathbf{h}_{22}}{1 + z_1^2} \right) \\ & \text{subject to} && \mathbf{h}_{12}^T \mathbf{S}_1 \mathbf{h}_{12} = z_1^2, \quad \mathbf{h}_{21}^T \mathbf{S}_2 \mathbf{h}_{21} = z_2^2, \\ & && \text{tr}(\mathbf{S}_i) \leq P_i, \quad \mathbf{S}_i \succeq 0, \quad i = 1, 2. \end{aligned}$$

New Convex Optimization Problem

$$\begin{aligned}
 & \max \quad \mathbf{h}_{ii}^T \mathbf{S}_i \mathbf{h}_{ii} \\
 & \text{subject to} \quad \mathbf{h}_{ij}^T \mathbf{S}_i \mathbf{h}_{ij} \leq z_{ij}^2, \\
 & \quad \text{tr}(\mathbf{S}_i) \leq P_i, \quad \mathbf{S}_i \succeq 0 \\
 & \quad i, j = 1, \dots, m, \quad i \neq j,
 \end{aligned}$$

- Preselect z_{ji}^2 as the interference power caused by the j th transmitter to the i th receiver.
- Maximize useful signal power with the interference constraint.
- Obtain the optimal covariance matrix for user i .

Comparison of the Complexity

- Old problem

$$\prod_{i=1}^m (\text{num of } \mu_i) \text{ non-convex problems}$$

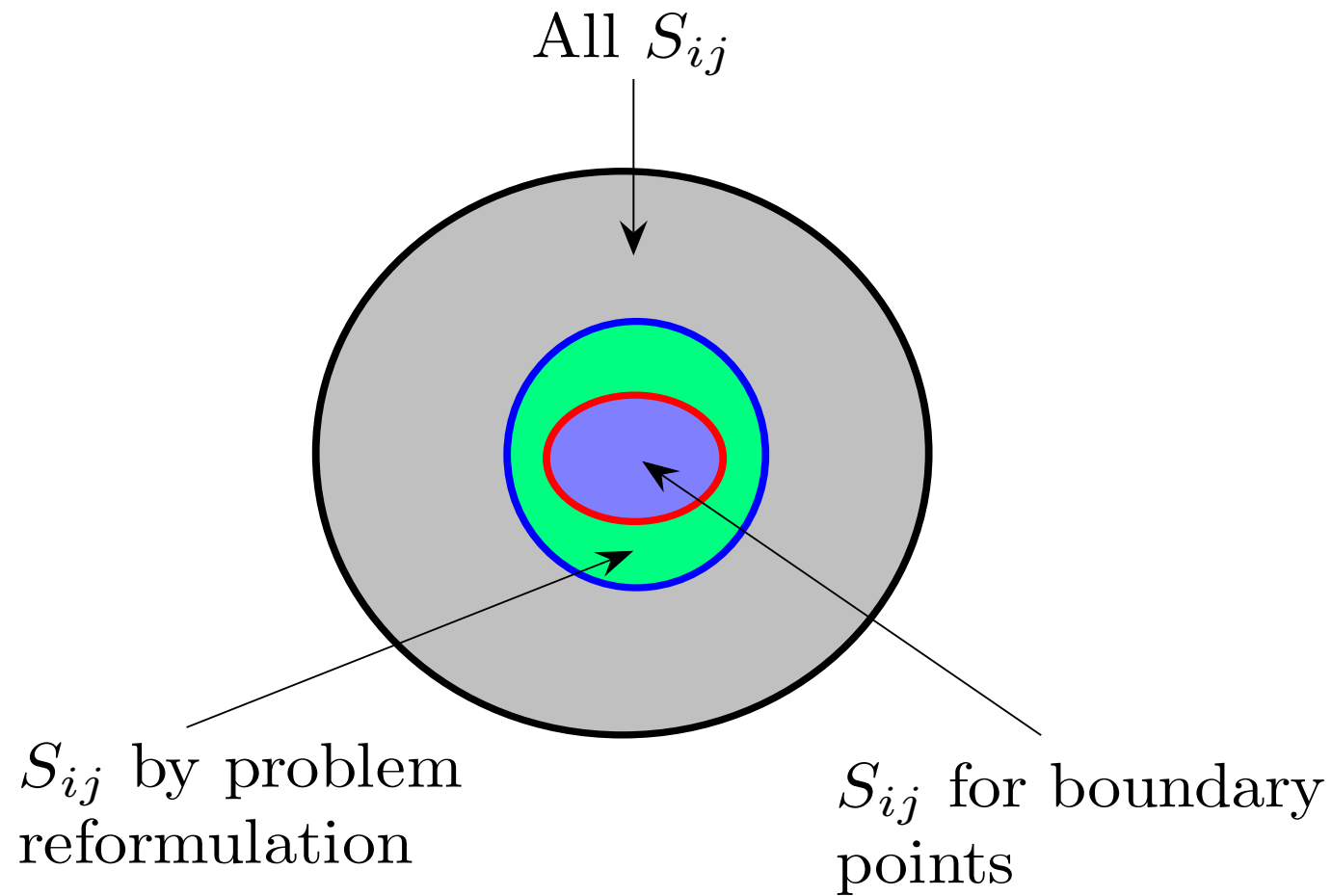
- New problem

$$\sum_{i=1}^m \sum_{j=1, j \neq i}^m (\text{num of } z_{ij}) \text{ convex problems}$$

- Advantage of problem reformulation

- Non-convex \Rightarrow convex
- Decoupling reduces complexity.

Covariance Matrices



Matrix Optimization Lemma

If

$$\mathbf{K} = \begin{bmatrix} \mathbf{K}_{11} & \mathbf{K}_{21}^T \\ \mathbf{K}_{21} & \mathbf{K}_{22} \end{bmatrix} \succeq \mathbf{0}, \quad \text{tr}(\mathbf{K}) \leq P$$

and $\mathbf{K}_{11} \succeq \mathbf{0}$ is a preselected sub-matrix, then

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix}^T \mathbf{K} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} \leq \left(\sqrt{\mathbf{x}^T \mathbf{K}_{11} \mathbf{x}} + \|\mathbf{y}\| \sqrt{P - \text{tr}(\mathbf{K}_{11})} \right)^2$$

and there exists \mathbf{K}^* that achieves the equality with

$$\text{rank}(\mathbf{K}^*) = \text{rank}(\mathbf{K}_{11}).$$

Problem Simplification

$$\begin{aligned}
 & \max \quad \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix}^T \mathbf{K} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \end{bmatrix} \\
 & \text{subject to} \quad h_i(\mathbf{K}_{11}) = 0, \quad g_j(\mathbf{K}_{11}) \leq 0 \\
 & \quad \quad \quad \text{tr}(\mathbf{K}) \leq P, \quad \mathbf{K} \succeq \mathbf{0},
 \end{aligned}$$

\Leftrightarrow

$$\begin{aligned}
 & \max \quad \left(\sqrt{\mathbf{x}^T \mathbf{K}_{11} \mathbf{x}} + \|\mathbf{y}\| \sqrt{P - \text{tr}(\mathbf{K}_{11})} \right)^2 \\
 & \text{subject to} \quad h_i(\mathbf{K}_{11}) = 0, \quad g_j(\mathbf{K}_{11}) \leq 0, \\
 & \quad \quad \quad \text{tr}(\mathbf{K}_{11}) \leq P, \quad \mathbf{K} \succeq \mathbf{0}.
 \end{aligned}$$

- Reduce the dimension of the matrix and vectors.

A Revisit of the Problem

$$\begin{aligned}
 & \max \quad \mathbf{h}_{mm}^T \mathbf{S}_m \mathbf{h}_{mm} \\
 & \text{subject to} \quad \mathbf{h}_{mj}^T \mathbf{S}_m \mathbf{h}_{mj} \leq z_{mj}^2, \quad j = 1, \dots, m-1 \\
 & \quad \text{tr}(\mathbf{S}_m) \leq P_m, \quad \mathbf{S}_m \succeq \mathbf{0}.
 \end{aligned}$$

\Rightarrow

$$\begin{aligned}
 & \max \quad \begin{bmatrix} \mathbf{h} \\ \hat{\mathbf{h}} \end{bmatrix}^T \tilde{\mathbf{S}} \begin{bmatrix} \mathbf{h} \\ \hat{\mathbf{h}} \end{bmatrix} \\
 & \text{subject to} \quad \begin{bmatrix} \mathbf{h}_j \\ \mathbf{0} \end{bmatrix}^T \tilde{\mathbf{S}} \begin{bmatrix} \mathbf{h}_j \\ \mathbf{0} \end{bmatrix} \leq z_{mj}^2, \quad j = 1, \dots, m-1, \\
 & \quad \text{tr}(\tilde{\mathbf{S}}) \leq P_m, \quad \tilde{\mathbf{S}} \succeq \mathbf{0},
 \end{aligned}$$

A Revisit of the Problem (continue)

\Rightarrow

$$\begin{aligned} & \max \quad \mathbf{h}^T \tilde{\mathbf{S}}_{11} \mathbf{h} \\ \text{subject to} \quad & \mathbf{h}_j^T \tilde{\mathbf{S}}_{11} \mathbf{h}_j \leq z_{mj}^2, \quad j = 1, \dots, m-1 \\ & \text{tr}(\tilde{\mathbf{S}}_{11}) \leq P_m, \quad \tilde{\mathbf{S}}_{11} \succeq 0, \end{aligned}$$

where

$$\begin{aligned} \dim(\mathbf{h}) = \dim(\mathbf{h}_j) &= \min \{t_m, m-1\} \triangleq \bar{m} \\ \tilde{\mathbf{S}} &= \begin{bmatrix} \tilde{\mathbf{S}}_{11} & \tilde{\mathbf{S}}_{21}^T \\ \tilde{\mathbf{S}}_{21} & \tilde{\mathbf{S}}_{22} \end{bmatrix}. \end{aligned}$$

Problem Simplification Procedure

- **Step 1:**

- Singular value decomposition

$$\mathbf{h}_{m1} = \mathbf{U}_1 \begin{bmatrix} \|\mathbf{h}_{m1}\| \\ \mathbf{0}_{(t_m-1) \times 1} \end{bmatrix}, \quad \text{where } \mathbf{U}_1^T \mathbf{U}_1 = \mathbf{I}$$

- then update

$$\mathbf{h}_{mj}^{(1)} = \mathbf{U}_1^T \mathbf{h}_{mj}, \quad j = 1, \dots, m.$$

$$\mathbf{S}_m^{(1)} = \mathbf{U}_1^T \mathbf{S}_m \mathbf{U}_1.$$

Problem Simplification Procedure (continue)

- **Step 2:**

$$\mathbf{h}_{m2}^{(1)} = \begin{bmatrix} \left(\mathbf{h}_{m2}^{(1)}\right)_1 \\ \left(\mathbf{h}_{m2}^{(1)}\right)_{2,\dots,t_m} \end{bmatrix} = \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \mathbf{U}_2 \end{bmatrix} \begin{bmatrix} \left(\mathbf{h}_{m2}^{(1)}\right)_1 \\ \left\| \left(\mathbf{h}_{m2}^{(1)}\right)_{2,\dots,t_m} \right\| \\ \mathbf{0}_{(t_m-2)\times 1} \end{bmatrix}, \quad \mathbf{U}_2^T \mathbf{U}_2 = \mathbf{I}$$

$$\mathbf{h}_{mj}^{(2)} = \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \mathbf{U}_2 \end{bmatrix}^T \mathbf{h}_{mj}^{(1)}, \quad j = 1, \dots, m.$$

$$\mathbf{S}_m^{(2)} = \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \mathbf{U}_2 \end{bmatrix}^T \mathbf{S}_m^{(1)} \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \mathbf{U}_2 \end{bmatrix}$$

Final State

$$\max \begin{bmatrix} \mathbf{h} \\ \widehat{\mathbf{h}} \end{bmatrix}^T \begin{bmatrix} \widetilde{\mathbf{S}}_{11} & \widetilde{\mathbf{S}}_{21}^T \\ \widetilde{\mathbf{S}}_{21} & \widetilde{\mathbf{S}}_{22} \end{bmatrix} \begin{bmatrix} \mathbf{h} \\ \widehat{\mathbf{h}} \end{bmatrix}$$

$$\text{subject to} \quad \begin{bmatrix} \mathbf{h}_j \\ \mathbf{0} \end{bmatrix}^T \begin{bmatrix} \widetilde{\mathbf{S}}_{11} & \widetilde{\mathbf{S}}_{21}^T \\ \widetilde{\mathbf{S}}_{21} & \widetilde{\mathbf{S}}_{22} \end{bmatrix} \begin{bmatrix} \mathbf{h}_j \\ \mathbf{0} \end{bmatrix} \leq z_{mj}^2, \quad j = 1, \dots, \bar{m},$$

$$\text{tr} \left(\begin{bmatrix} \widetilde{\mathbf{S}}_{11} & \widetilde{\mathbf{S}}_{21}^T \\ \widetilde{\mathbf{S}}_{21} & \widetilde{\mathbf{S}}_{22} \end{bmatrix} \right) \leq P_m, \quad \begin{bmatrix} \widetilde{\mathbf{S}}_{11} & \widetilde{\mathbf{S}}_{21}^T \\ \widetilde{\mathbf{S}}_{21} & \widetilde{\mathbf{S}}_{22} \end{bmatrix} \succeq \mathbf{0},$$

Optimality of Beamforming

• Theorem:

There exist an $\mathbf{S}^* \succeq \mathbf{0}$ and $\text{rank}(\mathbf{S}^*) \leq 1$ that is optimal for

$$\begin{aligned} \max \quad & \mathbf{h}_{mm}^T \mathbf{S}_m \mathbf{h}_{mm} \\ \text{subject to} \quad & \mathbf{h}_{mj}^T \mathbf{S}_m \mathbf{h}_{mj} \leq z_{mj}^2, \quad j = 1, \dots, m-1 \\ & \text{tr}(\mathbf{S}_m) \leq P_m, \quad \mathbf{S}_m \succeq \mathbf{0}. \end{aligned}$$

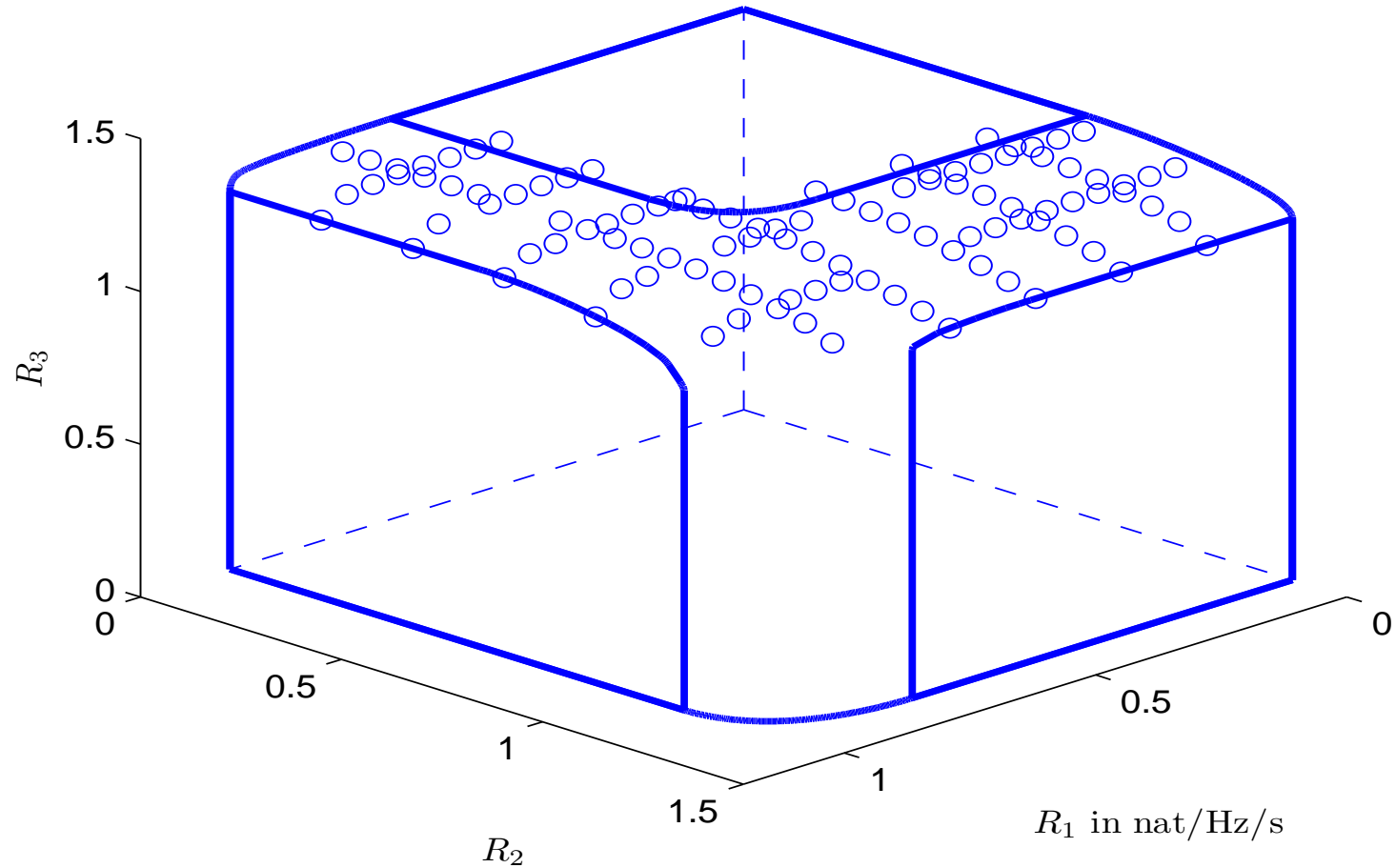
• Proof:

- Karush-Kuhn-Tucker conditions.
- Extension of Sylvester's Law of Inertia.

Conclusion

For an m -user MISO interference, the boundary points of the single-user detection rate region can be achieved by restricting each transmitter to implementing **beamforming**.

Numerical Example



The convex hull of the SUD rate regions of a three-user MISO IC.