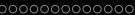


Generalized Degrees of Freedom of K user Symmetric Gaussian MIMO Interference Channel

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Outline

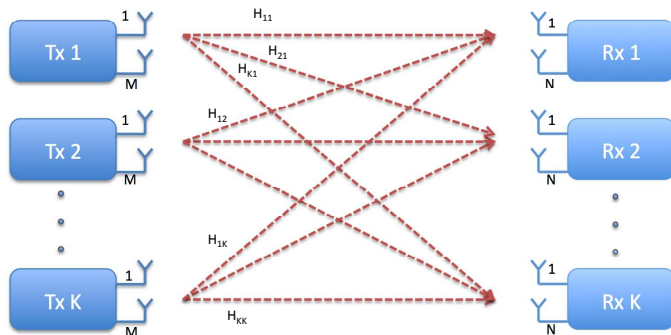
- Preliminaries
- Outer bound
- Inner bound
- Results
- Conclusion

Interference in wireless networks

- Shannon provides the basis for modern day communication system
- In point-to-point communication system noise is the primary concern
- Wireless networks are interference limited rather than noise limited



Interference Channel (IC)



- \mathbf{H}_{ij} : channel from j th transmitter to i th receiver
- M and N : antenna at transmitter and receiver respectively

Problem statement

- Multiple antennas help to mitigate the effect of interference e.g: When $N \geq KM$, ZF - receiving is sufficient to achieve the interference free GDOF
- When $N < KM$, trivial techniques are found to be sub-optimal
- **Focus of this work:** To characterize GDOF of K user symmetric MIMO Gaussian IC



Contributions

- A new outer bound is derived for the MIMO-IC
 - Cooperation
 - Providing noisy side information
- Inner bound is derived for the symmetric MIMO-IC as a combination of
 - Han-Kobayashi (HK) scheme
 - Interference Alignment (IA)
 - Treating interference as noise
 - Zero - Forcing receiving
- HK scheme is extended to multiuser MIMO scenario
- Interplay between the HK and IA schemes is explored

Outer bound

Fano's Inequality:

For any estimator \hat{X} such that $X \rightarrow Y \rightarrow \hat{X}$, with $P_e = \Pr\{X \neq \hat{X}\}$, we have:

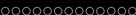
$$H(X|Y) \leq H(X|\hat{X}) \leq H(P_e) + P_e \log |X|$$

Weaker Form:

$$H(X|Y) \leq 1 + P_e \log |X|$$

- How to use:

$$\begin{aligned} nR_1 &= H(W_1) \\ &= I(W_1; Y_1^n) + H(W_1|Y_1^n) \\ &\leq I(W_1; Y_1^n) + n\epsilon_n \text{ (Fano's inequality)} \\ &\leq I(X_1^n; Y_1^n) + n\epsilon_n \text{ (Data processing inequality)} \end{aligned}$$



Outer bound based on cooperation

- Cooperation does not hurt capacity
- Outer bound is derived for a modified system
- Different possible ways of cooperation is taken in to account

Outer bound: cooperation

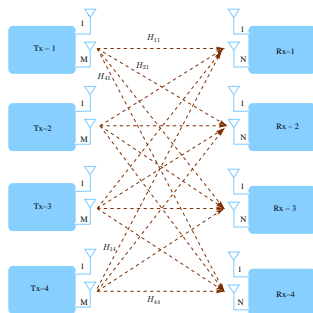
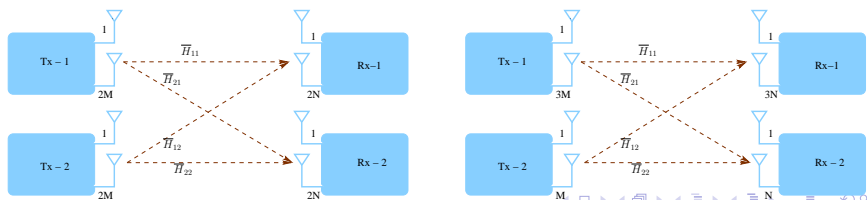
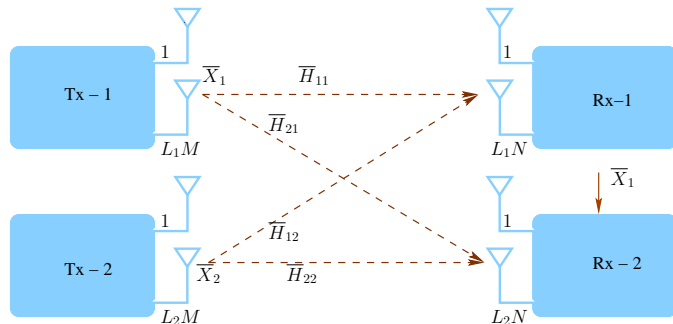


Figure: Four user Gaussian IC



Outer bound: cooperation



- L_1 and L_2 : number of users in group - 1 and group -2
- $\bar{H}_{ij} \in \mathbb{C}^{L_i N \times L_j M}$, \bar{X}_1 and \bar{X}_2 : two set of messages



Outer bound: cooperation

- System model:

$$\bar{\mathbf{Y}}_1 = \bar{\mathbf{H}}_{11}\bar{\mathbf{X}}_1 + \bar{\mathbf{H}}_{12}\bar{\mathbf{X}}_2 + \bar{\mathbf{Z}}_1,$$

$$\bar{\mathbf{Y}}_2 = \bar{\mathbf{H}}_{22}\bar{\mathbf{X}}_2 + \bar{\mathbf{Z}}_2,$$

where

$$\begin{aligned}\bar{\mathbf{Y}}_1 &\triangleq [\mathbf{Y}_1, \dots, \mathbf{Y}_{L_1}]^T, & \bar{\mathbf{Y}}_2 &\triangleq [\mathbf{Y}_{L_1+1}, \dots, \mathbf{Y}_L]^T, \\ \bar{\mathbf{X}}_1 &\triangleq [\mathbf{X}_1, \dots, \mathbf{X}_{L_1}]^T, & \bar{\mathbf{X}}_2 &\triangleq [\mathbf{X}_{L_1+1}, \dots, \mathbf{X}_L]^T, \\ \bar{\mathbf{Z}}_1 &\triangleq [\mathbf{Z}_1, \dots, \mathbf{Z}_{L_1}]^T & \text{and } \bar{\mathbf{Z}}_2 &\triangleq [\mathbf{Z}_{L_1+1}, \dots, \mathbf{Z}_L]^T.\end{aligned}$$

$$\bar{\mathbf{H}}_{11} = \text{blkdiag}(\mathbf{H}_{11} \mathbf{H}_{22} \dots \mathbf{H}_{L_1, L_1})$$

$$\bar{\mathbf{H}}_{22} = \text{blkdiag}(\mathbf{H}_{L_1+1, L_1+1} \mathbf{H}_{L_1+2, L_1+2} \dots \mathbf{H}_{L, L})$$

$$\bar{\mathbf{H}}_{12} = \begin{bmatrix} \mathbf{H}_{1, L_1+1} & \mathbf{H}_{1, L_1+2} & \cdots & \mathbf{H}_{1, L} \\ \mathbf{H}_{2, L_1+1} & \mathbf{H}_{2, L_1+2} & \cdots & \mathbf{H}_{2, L} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{H}_{L_1, L_1+1} & \mathbf{H}_{L_1, L_1+2} & \cdots & \mathbf{H}_{L_1, L} \end{bmatrix}, \quad L_1 + L_2 \leq K$$

Outer bound: cooperation

Theorem - 1:

The sum rate of the K -user Gaussian MIMO interference channel is upper bounded as follows:

$$\sum_{i=1}^L R_i \leq \log \left| \mathbf{I}_{L_1 N} + \bar{\mathbf{H}}_{11} \bar{\mathbf{P}}_1 \bar{\mathbf{H}}_{11}^H + \bar{\mathbf{H}}_{12} \bar{\mathbf{P}}_2 \bar{\mathbf{H}}_{12}^H \right| +$$

$$\log \left| \mathbf{I}_{L_2 N} + \bar{\mathbf{H}}_{22} \bar{\mathbf{P}}_2^{1/2} \left\{ \mathbf{I}_{L_2 M} + \bar{\mathbf{P}}_2^{1/2} \bar{\mathbf{H}}_{12}^H \bar{\mathbf{H}}_{12} \bar{\mathbf{P}}_2^{1/2} \right\}^{-1} \bar{\mathbf{P}}_2^{1/2} \bar{\mathbf{H}}_{22}^H \right| + \epsilon_n$$

where $L_1 + L_2 = L \leq K$, $0 \leq L_1 \leq K$, $0 \leq L_2 \leq K$,

\mathbf{I}_L : $L \times L$ identity matrix

Proof outline: cooperation

By using Fano's inequality, the sum rate of the modified system is upper bounded as given below:

$$\begin{aligned}
 & n \sum_{i=1}^L R_i - n\epsilon_n \\
 & \leq I(\bar{\mathbf{X}}_1^n, \bar{\mathbf{Y}}_1^n) + I(\bar{\mathbf{X}}_2^n; \bar{\mathbf{Y}}_2^n), \\
 & \leq I(\bar{\mathbf{X}}_1^n, \bar{\mathbf{Y}}_1^n) + I(\bar{\mathbf{X}}_2^n; \bar{\mathbf{Y}}_2^n, \bar{\mathbf{S}}^n), \text{ where } \bar{\mathbf{S}} = \bar{\mathbf{H}}_{12} \bar{\mathbf{X}}_2 + \bar{\mathbf{Z}}_1, \\
 & = h(\bar{\mathbf{Y}}_1^n) - h(\bar{\mathbf{S}}^n) + h(\bar{\mathbf{S}}^n) - h(\bar{\mathbf{Z}}_1^n) + h(\bar{\mathbf{Y}}_2^n | \bar{\mathbf{S}}^n) - h(\bar{\mathbf{Z}}_2^n),
 \end{aligned}$$

Proof outline: cooperation

Lemma

Let $\mathbf{x}^n = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n\}$ and $\mathbf{y}^n = \{\mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_n\}$ be two sequences of random vectors and let $\mathbf{x}^*, \mathbf{y}^*, \hat{\mathbf{x}}$ and $\hat{\mathbf{y}}$ be Gaussian vectors with covariance matrices satisfying

$$\text{Cov} \begin{bmatrix} \hat{\mathbf{x}} \\ \hat{\mathbf{y}} \end{bmatrix} = \frac{1}{n} \sum_{i=1}^n \text{Cov} \begin{bmatrix} \mathbf{x}_i \\ \mathbf{y}_i \end{bmatrix} \preceq \text{Cov} \begin{bmatrix} \mathbf{x}^* \\ \mathbf{y}^* \end{bmatrix},$$

then we get the following bound

$$h(\mathbf{x}^n) \leq nh(\hat{\mathbf{x}}) \leq nh(\mathbf{x}^*),$$

$$h(\mathbf{y}^n | \mathbf{x}^n) \leq nh(\hat{\mathbf{y}} | \hat{\mathbf{x}}) \leq nh(\mathbf{y}^* | \mathbf{x}^*).$$

Proof outline: cooperation

- Sum rate is bounded as:

$$\sum_{i=1}^L R_i - \epsilon_n \leq h(\bar{\mathbf{Y}}_1^*) - h(\bar{\mathbf{Z}}_1) + h(\bar{\mathbf{Y}}_2^* | \bar{\mathbf{S}}^*) - h(\bar{\mathbf{Z}}_2)$$

- It can be shown that:

$$h(\bar{\mathbf{Y}}_1^*) = \log \left| \pi e \left[\mathbf{I}_{L_1 N} + \bar{\mathbf{H}}_{11} \bar{\mathbf{P}}_1 \bar{\mathbf{H}}_{11}^H + \bar{\mathbf{H}}_{12} \bar{\mathbf{P}}_2 \bar{\mathbf{H}}_{12}^H \right] \right|$$

$$h(\bar{\mathbf{Y}}_2^* | \bar{\mathbf{S}}^*) = \log \left| \pi e \left[\mathbf{I}_{L_2 N} + \bar{\mathbf{H}}_{22} \bar{\mathbf{P}}_2^{1/2} \left\{ \mathbf{I}_{L_2 M} + \bar{\mathbf{P}}_2^{1/2} \bar{\mathbf{H}}_{12}^H \bar{\mathbf{H}}_{12} \bar{\mathbf{P}}_2^{1/2} \right\}^{-1} \bar{\mathbf{P}}_2^{1/2} \bar{\mathbf{H}}_{22}^H \right] \right|.$$

- Need to minimize sum rate over all possible values of L_1 and L_2 .
- Difficult task !

Outer bound: cooperation

Lemma

In the symmetric case, the GDOF of the K user Gaussian MIMO-IC $M \leq N$ is upper bounded as follows:

- 1 When $0 \leq \alpha \leq 1$:

$$d_i(\alpha) \leq \min_{L_1, L_2} \frac{1}{L} \left[L_1 M + \min \{r, L_1(N - M)\} \alpha + (L_2 M - r)^+ + \min \{r, L_2 N - (L_2 M - r)^+\} (1 - \alpha) \right],$$

- 2 When $\alpha > 1$:

$$d_i(\alpha) \leq \min_{L_1, L_2} \frac{1}{L} \left[r \alpha + \min \{L_1 M, L_1 N - r\} + (L_2 M - r)^+ \right],$$

where $r = \min \{L_2 M, L_1 N\}$.

Outer bound: Noisy side information

- Noisy side information is provided
- Define the following quantity:

$$\mathbf{S}_{j,\mathbf{B}} = \sum_{i \in \mathbf{B}} \mathbf{H}_{ji} \mathbf{X}_i + \mathbf{Z}_j$$

where $\mathbf{B} \subseteq \{1, 2, \dots, K\}$ is the set of users.

- Consider first and third user:

$$\begin{aligned}
 & nR_1 + nR_3 - n\epsilon_n \\
 & \leq I(\mathbf{X}_1^n; \mathbf{Y}_1^n) + I(\mathbf{X}_3^n; \mathbf{Y}_3^n) \\
 & \leq I(\mathbf{X}_1^n; \mathbf{Y}_1^n, \mathbf{S}_{2,1}^n) + I(\mathbf{X}_3^n; \mathbf{Y}_3^n, \mathbf{S}_{4,3}^n) \\
 & = h(\mathbf{Y}_1^n | \mathbf{S}_{2,1}^n) + h(\mathbf{Y}_3^n | \mathbf{S}_{4,3}^n) + \underbrace{h(\mathbf{S}_{2,1}^n) + h(\mathbf{S}_{4,3}^n)}_{\text{unwanted terms}} \\
 & \quad - \underbrace{h(\mathbf{S}_{1,\{2,3,4\}}^n) - h(\mathbf{S}_{3,\{1,2,4\}}^n)}_{\text{unwanted terms}} - h(\mathbf{Z}_1^n) - h(\mathbf{Z}_3^n)
 \end{aligned}$$

Outer bound: Noisy side information

- Consider 2nd and 4th user:

$$\begin{aligned}
 & nR_2 + nR_4 - n\epsilon_n \\
 & \leq I(\mathbf{X}_2^n; \mathbf{Y}_2^n, \mathbf{S}_{1,2}^n) + I(\mathbf{X}_4^n; \mathbf{Y}_4^n, \mathbf{S}_{3,4}^n) \\
 & = h(\mathbf{Y}_2^n | \mathbf{S}_{1,2}^n) + h(\mathbf{Y}_4^n | \mathbf{S}_{3,4}^n) + \underbrace{h(\mathbf{S}_{1,2}^n) + h(\mathbf{S}_{3,4}^n)}_{\text{unwanted terms}} \\
 & \quad - \underbrace{h(\mathbf{S}_{2,\{1,3,4\}}^n) - h(\mathbf{S}_{4,\{1,2,3\}}^n)}_{\text{unwanted terms}} - h(\mathbf{Z}_2^n) - h(\mathbf{Z}_4^n)
 \end{aligned}$$

- Summing and by conditioning further:

$$\begin{aligned}
 \sum_{i=1}^4 R_i & \leq h(\mathbf{Y}_1^n | \mathbf{S}_{2,1}^n) + h(\mathbf{Y}_3^n | \mathbf{S}_{4,3}^n) + h(\mathbf{Y}_2^n | \mathbf{S}_{1,2}^n) + h(\mathbf{Y}_4^n | \mathbf{S}_{3,4}^n) \\
 & \quad + \sum_{i=1}^4 h(\mathbf{Z}_i^n) + n\epsilon_n
 \end{aligned}$$

Outer bound: Noisy side information

Theorem

The sum rate of the K user Gaussian MIMO-IC is upper bounded as follows:

① When K is even:

$$\sum_{i=1}^K R_i \leq \sum_{i \text{ odd}} \log |\mathbf{I}_{N_i} + \sum_{j=1, j \neq i}^K \mathbf{H}_{ij} \mathbf{P}_j \mathbf{H}_{ij}^H + \mathbf{H}_{ii} \mathbf{P}_i^{1/2} \{ \mathbf{I}_{M_i} + \mathbf{P}_i^{1/2} \mathbf{H}_{i+1,i}^H \mathbf{H}_{i+1,i} \mathbf{P}_i^{1/2} \}^{-1} \mathbf{P}_i^{1/2} \mathbf{H}_{ii}^H | +$$

$$\sum_{i \text{ even}} \log |\mathbf{I}_{N_i} + \sum_{j=1, j \neq i}^K \mathbf{H}_{ij} \mathbf{P}_j \mathbf{H}_{ij}^H + \mathbf{H}_{ii} \mathbf{P}_i^{1/2} \{ \mathbf{I}_{M_i} + \mathbf{P}_i^{1/2} \mathbf{H}_{i-1,i}^H \mathbf{H}_{i-1,i} \mathbf{P}_i^{1/2} \}^{-1} \mathbf{P}_i^{1/2} \mathbf{H}_{ii}^H | + \epsilon_n$$

② When K is odd:

$$R_1 + 2 \sum_{i=2}^{K-1} R_i + R_K$$

$$\leq \sum_{i=1}^{K-1} \log |\mathbf{I}_{N_i} + \sum_{j=1, j \neq i}^K \mathbf{H}_{ij} \mathbf{P}_j \mathbf{H}_{ij}^H + \mathbf{H}_{ii} \mathbf{P}_i^{1/2} \left(\mathbf{I}_{M_i} + \mathbf{P}_i^{1/2} \mathbf{H}_{i+1,i}^H \mathbf{H}_{i+1,i} \mathbf{P}_i^{1/2} \right)^{-1} \mathbf{P}_i^{1/2} \mathbf{H}_{ii}^H | +$$

$$\sum_{i=2}^K \log |\mathbf{I}_{N_i} + \sum_{j=1, j \neq i}^K \mathbf{H}_{ij} \mathbf{P}_j \mathbf{H}_{ij}^H + \mathbf{H}_{ii} \mathbf{P}_i^{1/2} \left(\mathbf{I}_{M_i} + \mathbf{P}_i^{1/2} \mathbf{H}_{i-1,i}^H \mathbf{H}_{i-1,i} \mathbf{P}_i^{1/2} \right)^{-1} \mathbf{P}_i^{1/2} \mathbf{H}_{ii}^H | + \epsilon_n$$

Outer bound: Noisy side information

Lemma

The GDOF of the K user MIMO-IC in the symmetric case is upper bounded as follows:

$$d_j(\alpha) \leq \begin{cases} M(1 - \alpha) + \min \{ \min \{ N, (K - 1)M \}, N - M \} \alpha & \text{if } 0 \leq \alpha \leq \frac{1}{2} \\ \min \{ N, (K - 1)M \} \alpha + \min \{ M, N - \min \{ N, (K - 1)M \} \} (1 - \alpha) & \text{if } \frac{1}{2} \leq \alpha \leq 1 \\ \min \{ N, (K - 1)M \} & \text{if } \alpha \geq 1 \end{cases}$$

when $M \leq N$.

Outer bound: Noisy side information

Theorem

The sum rate of the K user Gaussian MIMO-IC is upper bounded as follows:

$$\begin{aligned}
 R_1 + \sum_{i=2}^{K-1} R_i + R_K &\leq \log \left| \mathbf{I}_{N_1} + \sum_{j=2}^K \mathbf{H}_{1j} \mathbf{P}_j \mathbf{H}_{1j}^H + \mathbf{H}_{11} \mathbf{P}_1^{1/2} \left\{ \mathbf{I}_{M_1} + \mathbf{P}_1^{1/2} \mathbf{H}_{K1}^H \mathbf{H}_{K1} \mathbf{P}_1^{1/2} \right\}^{-1} \mathbf{P}_1^{1/2} \mathbf{H}_{11}^H \right| + \\
 &\quad \sum_{i=2}^{K-1} \log \left| \mathbf{I}_{N_i} + \bar{\mathbf{H}}_{i1} \bar{\mathbf{P}}_{i1}^{1/2} \left\{ \mathbf{I}_{M_{r_i}} + \bar{\mathbf{P}}_{i1}^{1/2} \bar{\mathbf{H}}_{K1}^H \bar{\mathbf{H}}_{K1} \bar{\mathbf{P}}_{i1}^{1/2} \right\}^{-1} \bar{\mathbf{P}}_{i1}^{1/2} \bar{\mathbf{H}}_{i1}^H + \right. \\
 &\quad \left. \bar{\mathbf{H}}_{i,i+1} \bar{\mathbf{P}}_{i2}^{1/2} \left\{ \mathbf{I}_{M_{s_i}} + \bar{\mathbf{P}}_{i2}^{1/2} \bar{\mathbf{H}}_{1,i+1}^H \bar{\mathbf{H}}_{1,i+1} \bar{\mathbf{P}}_{i2}^{1/2} \right\}^{-1} \bar{\mathbf{P}}_{i2}^{1/2} \bar{\mathbf{H}}_{i,i+1}^H \right| + \\
 &\quad \sum_{i=2}^{K-1} \log \left| \mathbf{I}_{N_i} + \bar{\mathbf{H}}_{iK} \bar{\mathbf{P}}_{i3}^{1/2} \left\{ \mathbf{I}_{M_{r_i}} + \bar{\mathbf{P}}_{i3}^{1/2} \bar{\mathbf{H}}_{1i}^H \bar{\mathbf{H}}_{1i} \bar{\mathbf{P}}_{i3}^{1/2} \right\}^{-1} \bar{\mathbf{P}}_{i3}^{1/2} \bar{\mathbf{H}}_{iK}^H + \right. \\
 &\quad \left. \bar{\mathbf{H}}_{i,K-1} \bar{\mathbf{P}}_{i4}^{1/2} \left\{ \mathbf{I}_{M_{s_i}} + \bar{\mathbf{P}}_{i4}^{1/2} \bar{\mathbf{H}}_{K,i+1}^H \bar{\mathbf{H}}_{K,i+1} \bar{\mathbf{P}}_{i4}^{1/2} \right\}^{-1} \bar{\mathbf{P}}_{i4}^{1/2} \bar{\mathbf{H}}_{i,K-1}^H \right| + \\
 &\quad \log \left| \mathbf{I}_{N_K} + \sum_{j=1}^{K-1} \mathbf{H}_{Kj} \mathbf{P}_j \mathbf{H}_{Kj}^H + \mathbf{H}_{KK} \mathbf{P}_K^{1/2} \left\{ \mathbf{I}_{M_K} + \mathbf{P}_K^{1/2} \mathbf{H}_{1K}^H \mathbf{H}_{1K} \mathbf{P}_K^{1/2} \right\}^{-1} \mathbf{P}_K^{1/2} \mathbf{H}_{KK}^H \right| + \epsilon_n
 \end{aligned}$$

where $M_{r_i} = \sum_{j=1}^i M_j$ and $M_{s_i} = \sum_{j=i+1}^K M_j$

Outer bound: Noisy side information

- Variables are defined as follows:

$$\begin{aligned} \bar{\mathbf{H}}_{i1} &= [\mathbf{H}_{i1} \quad \mathbf{H}_{i2} \quad \dots \quad \mathbf{H}_{ii}], & \bar{\mathbf{H}}_{i,i+1} &= [\mathbf{H}_{i,i+1} \quad \mathbf{H}_{i,i+2} \quad \dots \quad \mathbf{H}_{iK}], \\ \bar{\mathbf{H}}_{Ki} &= [\mathbf{H}_{K1} \quad \mathbf{H}_{K2} \quad \dots \quad \mathbf{H}_{Ki}], & \bar{\mathbf{H}}_{1,i+1} &= [\mathbf{H}_{1,i+1} \quad \mathbf{H}_{1,i+2} \quad \dots \quad \mathbf{H}_{1K}], \\ \bar{\mathbf{H}}_{1i} &= [\mathbf{H}_{1K} \quad \mathbf{H}_{12} \quad \dots \quad \mathbf{H}_{1i}], & \bar{\mathbf{H}}_{K,i+1} &= [\mathbf{H}_{K1} \quad \mathbf{H}_{K,i+1} \quad \dots \quad \mathbf{H}_{K,K-1}], \\ \bar{\mathbf{H}}_{iK} &= [\mathbf{H}_{iK} \quad \mathbf{H}_{i2} \quad \dots \quad \mathbf{H}_{ii}], & \bar{\mathbf{H}}_{i,K-1} &= [\mathbf{H}_{i1} \quad \mathbf{H}_{i,i+1} \quad \dots \quad \mathbf{H}_{i,K-1}] \end{aligned}$$

$$\begin{aligned} \bar{\mathbf{P}}_{i1} &= \text{blockdiag}(\mathbf{P}_1 \mathbf{P}_2 \dots \mathbf{P}_i), & \bar{\mathbf{P}}_{i2} &= \text{blockdiag}(\mathbf{P}_{i+2} \mathbf{P}_{i+3} \dots \mathbf{P}_K) \\ \bar{\mathbf{P}}_{i3} &= \text{blockdiag}(\mathbf{P}_K \mathbf{P}_2 \dots \mathbf{P}_i) & \text{and } \bar{\mathbf{P}}_{i4} &= \text{blockdiag}(\mathbf{P}_1 \mathbf{P}_{i+1} \dots \mathbf{P}_{K-1}). \end{aligned}$$

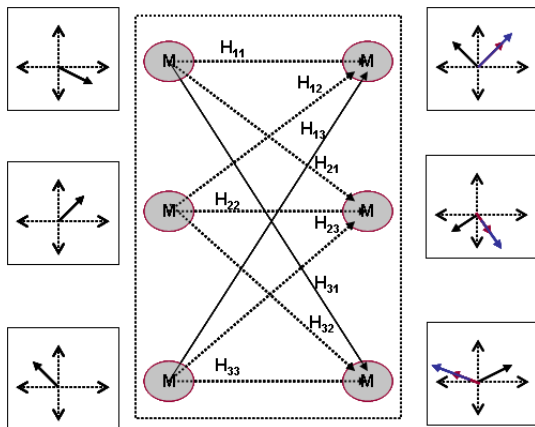
- Extension of SIMO-IC outer bound to MIMO-IC

Inner bound

- Inner bound is derived for the symmetric MIMO interference channel as a combination of
 - Han-Kobayashi (HK) scheme
 - Interference Alignment (IA)
 - Treating interference as noise
 - Zero - Forcing receiving
- Interplay between the HK and IA schemes is explored



Interference alignment(IA)



- Idea of IA for IC originated in the seminal work by Cadembe
- For MIMO IC, DOF achieved by IA:

$$d_j = \frac{MN}{M+N}, \quad KM > N \quad (2)$$

- Requires global channel knowledge
- Relative strength between signal and interference does not matter

Han-Kobayashi (HK) scheme

- Based on the idea of splitting message in to two parts:
 - 1 Private part
 - 2 Public part
- A simple HK scheme proposed by ETW: achieves capacity with in 1 bit (two user IC)
- Different variants of HK - scheme has been proposed

Extension of HK - scheme

- HK scheme is extended to K user MIMO IC for symmetric case
- Following interference regime are considered:
 - ① Strong interference case ($\alpha > 1$)
 - ② Moderate interference case ($\frac{1}{2} \leq \alpha \leq 1$)
 - ③ Weak interference case ($0 \leq \alpha \leq \frac{1}{2}$)

HK Scheme: Strong interference case

- Every receiver tries to decode the unintended messages along with the intended one
- There is no private part
- K - user MAC channel is formed at every receiver
- Achievable rate region: intersection of K - MAC regions

HK Scheme: Strong interference case

Theorem

The following GDOD is achievable in case of K user Gaussian MIMO-IC:

- 1 When $\frac{N}{M} < K \leq \frac{N}{M} + 1$:

$$d_j(\alpha) = \min \left\{ M, \frac{1}{K} [(K-1)M\alpha + N - (K-1)M] \right\}$$

- 2 When $K > \frac{N}{M} + 1$:

$$d_j(\alpha) = \min \left\{ M, \frac{\alpha N}{K} \right\}$$

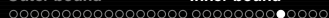


Proof outline: Strong interference case

- It is sufficient to consider any particular user due to symmetry of the problem
- For $S \subseteq \{1, 2, \dots, K\}$ a MAC channel is formed
- The rate that can be achieved at user - i :

$$\begin{aligned} \sum_{j \in S} R_j &= \log \left| \mathbf{I} + \rho \mathbf{H}_{11} \mathbf{H}_{11}^H + \rho^\alpha \sum_{j \neq 1} \mathbf{H}_{1j} \mathbf{H}_{1j}^H \right| \\ &= \alpha \min \{(K-1)M, N\} \log \rho + \\ &\quad \min \{M, N - \min \{(K-1)M, N\}\} \log \rho + \mathcal{O}(1) \end{aligned}$$

- Simplified based on the value of K , M and N



HK scheme: moderate interference case

- Message is split in to private and public part
- Private power is set such that it is received at the noise floor of the unintended receiver
- Both messages are encoded using Gaussian code book
- Decoding order:
 - While decoding the common message all private messages are treated as noise
 - Private message is decoded last: treat other user's private message as noise
- Rate achieved: $R_j = R_{p,j} + R_{c,j}$

HK scheme: weak interference case

Theorem

In case of Gaussian MIMO-IC, following GDOF is achievable:

$$d_j(\alpha) = M(1 - \alpha) + \frac{1}{K - 1}(N - M)$$

Comparison of different schemes

Kill it (suppression, orthogonalization)



Be friends (decode interference)

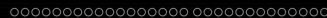


Ignore it (live with interference)



Let them fight with each other
(interference alignment)





Some insights

- Treating interference as noise is GDOF optimal when $M = N$ in weak interference case
- When $M < N$, splitting message into private and public part helps in weak interference regime
- When $K > \frac{N}{M} + 1$, a combination of IA and HK scheme performs better
- When $\frac{N}{M} < K \leq \frac{N}{M} + 1$, HK scheme is GDOF optimal
- Unlike two user IC, ZF - receiving is found to be GDOF optimal at $\alpha = 1$ when $\frac{N}{M} < K \leq \frac{N}{M} + 1$

