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Hybrid Precoding Using Long-term Channel Statistics for Massive MIMO Systems

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Contents

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- Proposed hybrid precoding techniques for massive MIMO
 - + Unconstrained case w/o phase shifter constraint
 - + Constrained case w/ phase shifter constraint
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Background: digital precoding vs. hybrid precoding



Key feature I) $M \le N$ (limited number of RF chains) Key feature 2) $|\mathbf{F}_{RF}(i,j)| = 1$ for i=1,...,N, and j=1,...,M (implemented with phase shifters)



Motivation: using long-term channel statistics



- 1) L. Liang, W. Xu, and X. Dong, "Low-complexity hybrid precoding in massive multiuser MIMO systems," IEEE Wireless Comm. Letters, Dec. 2014.
- 2) A.Alkhateeb, G. Leus, and R. Heath, "Limited feedback hybrid precoding for multi-user millimeter wave systems," IEEE Trans. on Wireless Comm., Nov. 2015.
- 3) F. Sohrabi and W.Yu, "Hybrid digital and analog beamforming design for large-scale antenna arrays," IEEE JSTSP, Apr. 2016.



Overview of proposed techniques





[Proposed I] Unconstrained case: Prior work

Prior work

+ \mathbf{F}_{RF} : Each column of \mathbf{F}_{RF} is a dominant eigenvector of each user's \mathbf{R}_k matrix.

Channel model		
$\mathbf{H} = egin{bmatrix} \mathbf{h}_1 \ dots \ \mathbf{h}_K \end{bmatrix} = egin{bmatrix} \mathbf{h}_{w,1} \mathbf{R}_1^{rac{1}{2}} \ dots \ \mathbf{h}_{w,K} \mathbf{R}_K^{rac{1}{2}} \end{bmatrix}$		
$\mathbf{h}_{w,k}: 1 imes N$, $\mathcal{CN}(0, \mathbf{I})$		
$\mathbf{R}_k = \mathbb{E}\left[\mathbf{h}_k^H \mathbf{h}_k ight] = \mathbf{V}_k \mathbf{\Sigma}_k \mathbf{V}_k^H$		
(spatial channel covariance matrix of user k)		

Hybrid precoding design [Prior work ¹⁻²]

$$\mathbf{F}_{\mathrm{RF}} = \begin{bmatrix} \mathbf{v}_{1,max} & \cdots & \mathbf{v}_{K,max} \end{bmatrix}$$

where $\mathbf{v}_{k,max}$: dominant eigen vector of \mathbf{R}_k

$$\mathbf{F}_{\rm BB} = \left(\mathbf{F}_{\rm RF}^* \mathbf{H} \mathbf{H}^* \mathbf{F}_{\rm RF} + \frac{K\sigma^2}{P_{\rm tx}} \mathbf{I}_M\right)^{-1} \mathbf{F}_{\rm RF}^* \mathbf{H}$$

[Note] The index term [t] is omitted for simplicity. Note that \mathbf{h}_k , $\mathbf{h}_{w,k}$, and \mathbf{F}_{BB} depend on t, and \mathbf{R}_k , $\mathbf{v}_{k,max}$, and \mathbf{F}_{RF} do not depend on t.

- Key idea: F_{RF} maximizes the long-term power of each user's desired signal in the analog part.
- Shortcomings: Does not consider interference and can be used only when M = K.
- L. Liang, Y. Dai, W. Xu, and X. Dong, "How to approach zero-forcing under RF chain limitations in large mmWave multiuser systems?," *IEEE/CIC*, Oct. 2014
 A. Alkhateeb, G. Leus, and R.W. Heath, "Multilayer precoding for full-dimensional massive MIMO systems," *ASILOMAR*, Nov. 2014.



[Proposed I] Unconstrained case: Proposed (1/5)

Proposed

- \mathbf{F}_{RF} : The columns of \mathbf{F}_{RF} constitute a subspace that maximizes the performance of the baseband precoder
 - Note: Each column is not assigned to a specific user.

Definition of performance : Asymptotic average SLNR ¹⁾ (a deterministic value as $N \rightarrow \infty$)

I) SLNR: signal-to-leakage-plus-noise ratio

	Main role of \mathbf{F}_{RF}	Remark
Prior work	Maximize each user's long-term desired power in the analog part	Can be used only when $M = K$
Proposed approach	Maximize the performance of ${f F}_{BB}$ in the analog part (i.e., help ${f F}_{BB}$ to work well)	Can be used when $M \ge K$

- Key idea: \mathbf{F}_{RF} helps \mathbf{F}_{BB} (R-ZF) to minimize the inter-user interference.
- Can be applied to the case of $M \ge K$



[Proposed I] Unconstrained case: Proposed (2/5)

I) System model (R-ZF is used in the baseband)

 $\mathbf{y} = \mathbf{H}^* \mathbf{x} + \mathbf{n} = \mathbf{H}^* \mathbf{F}_{\mathrm{RF}} \mathbf{F}_{\mathrm{BB}} \mathbf{P} \mathbf{s} + \mathbf{n}$

$$\mathbf{F}_{\mathrm{BB}} = \left(\mathbf{F}_{\mathrm{RF}}^{*}\mathbf{H}\mathbf{H}^{*}\mathbf{F}_{\mathrm{RF}} + \frac{K\sigma^{2}}{P_{\mathrm{tx}}}\mathbf{I}_{M}\right)^{-1}\mathbf{F}_{\mathrm{RF}}^{*}\mathbf{H} = \begin{bmatrix}\mathbf{f}_{\mathrm{bb},1} & \cdots & \mathbf{f}_{\mathrm{bb},K}\end{bmatrix}$$

$$\mathbf{f}_{ ext{bb},k} = \left(\mathbf{F}_{ ext{RF}}^{*}\mathbf{H}\mathbf{H}^{*}\mathbf{F}_{ ext{RF}} + rac{K\sigma^{2}}{P_{ ext{tx}}}\mathbf{I}_{M}
ight)^{-1}\mathbf{F}_{ ext{RF}}^{*}\mathbf{h}_{k}$$

$$\mathbf{P} = ext{diag} \left(egin{bmatrix} p_1 & \cdots & p_K \end{bmatrix}
ight)$$

$$p_k = rac{\sqrt{P_{tx}}}{\sqrt{K} \|\mathbf{F}_{ ext{RF}} \mathbf{f}_{ ext{bb},k}\|}$$

$$= \sqrt{rac{P_{tx}}{K \mathbf{h}_k^* \mathbf{F}_{ ext{RF}} \mathbf{W} \mathbf{F}_{ ext{RF}}^* \mathbf{F}_{ ext{RF}} \mathbf{W} \mathbf{F}_{ ext{RF}}^* \mathbf{h}_k}$$

2) Let each column of \mathbf{F}_{RF} be a linear combination of orthonormal bases $\{\mathbf{v}_1, \dots, \mathbf{v}_M\}$ (subspace)

$$\mathbf{F}_{\mathrm{RF}} = \mathbf{V}\mathbf{A}$$

where $\mathbf{A} \in \mathcal{C}^{M \times M}$ is an $M \times M$ matrix, and
 $\mathbf{V} = \begin{bmatrix} \mathbf{v}_1 & \cdots & \mathbf{v}_M \end{bmatrix} \in \mathcal{U}^{N \times M} (\mathbf{V}^H \mathbf{V} = \mathbf{I}_M)$

3) Proposition

If V and P_{tx} are given, SLNR is maximized when A is unitary, which indicates that \mathbf{F}_{RF} must be a semiunitary matrix to maximize SLNR.



[Proposed I] Unconstrained case: Proposed (4/5)

4) If A is unitary, the SLNR becomes

$$\mathrm{SLNR}_{k} = \frac{\mathbf{h}_{k}^{*} \mathbf{V} \left(\mathbf{V}^{*} \mathbf{H} \mathbf{H}^{*} \mathbf{V} + \frac{K\sigma^{2}}{P_{\mathrm{tx}}} \mathbf{I}_{M} \right)^{-1} \mathbf{V}^{*} \mathbf{h}_{k}}{1 - \mathbf{h}_{k}^{*} \mathbf{V} \left(\mathbf{V}^{*} \mathbf{H} \mathbf{H}^{*} \mathbf{V} + \frac{K\sigma^{2}}{P_{\mathrm{tx}}} \mathbf{I}_{M} \right)^{-1} \mathbf{V}^{*} \mathbf{h}_{k}} = \mathbf{h}_{k}^{*} \mathbf{V} \left(\mathbf{V}^{*} \left(\sum_{i \neq k}^{K} \mathbf{h}_{i} \mathbf{h}_{i}^{*} \right) \mathbf{V} + \frac{K\sigma^{2}}{P_{\mathrm{tx}}} \mathbf{I}_{M} \right)^{-1} \mathbf{V}^{*} \mathbf{h}_{k}}$$

5) Considering large antenna arrays, the SLNR converges to

$$\begin{aligned} \text{SLNR}_{k} &= N \mathbf{g}_{k}^{H} \mathbf{R}_{k}^{\frac{1}{2}} \mathbf{V} \left(N \sum_{i \neq k}^{K} \mathbf{V}^{*} \mathbf{R}_{i}^{\frac{1}{2}} \mathbf{g}_{i} \mathbf{g}_{i}^{*} \mathbf{R}_{i}^{\frac{1}{2}} \mathbf{V} + \frac{K \sigma^{2}}{P_{\text{tx}}} \mathbf{I}_{N} \right)^{-1} \mathbf{V}^{*} \mathbf{R}_{k}^{\frac{1}{2}} \mathbf{g}_{k} \\ & \xrightarrow{a.s.} \text{Tr} \left(\mathbf{V}^{*} \mathbf{R}_{k} \mathbf{V} \left(N \sum_{i=1}^{K} \mathbf{V}^{*} \mathbf{R}_{i}^{\frac{1}{2}} \mathbf{g}_{i} \mathbf{g}_{i}^{*} \mathbf{R}_{i}^{\frac{1}{2}} \mathbf{V} + \frac{K \sigma^{2}}{P_{\text{tx}}} \mathbf{I}_{N} \right)^{-1} \right)^{-1} \left[\xrightarrow{a.s.} \gamma_{k} \right] \\ & \text{where } \gamma_{1}, ..., \gamma_{K} \text{ are the unique nonnegative solution of} \\ & \gamma_{k} = \text{Tr} \left(\mathbf{V}^{*} \mathbf{R}_{k} \mathbf{V} \left(\left(\sum_{j=1}^{K} \frac{\mathbf{V}^{*} \mathbf{R}_{j} \mathbf{V}}{1 + \gamma_{j}} + \frac{K \sigma^{2}}{P_{\text{tx}}} \mathbf{I}_{N} \right) \right)^{-1} \right) \end{aligned}$$



[Proposed I] Unconstrained case: Proposed (5/5)

6) Optimization problem and its relaxation

$$\max \frac{1}{K} \sum_{k=1}^{K} \gamma_{k}, \text{ s.t. } \gamma_{k} = \operatorname{Tr} \left(\mathbf{V}^{*} \mathbf{R}_{k} \mathbf{V} \left(\sum_{j=1}^{K} \frac{\mathbf{V}^{*} \mathbf{R}_{j} \mathbf{V}}{1 + \gamma_{j}} + \frac{K}{\rho} \mathbf{I}_{M} \right)^{-1} \right), \forall k$$

$$\lim_{\mathbf{V} \in \mathbb{U}^{N \times M}} \left[\widehat{\gamma}_{1} = \dots = \gamma_{K} = \gamma = \frac{1}{K} \sum_{k=1}^{K} \gamma_{k} \right]$$

$$\lim_{\mathbf{V} \in \mathbb{U}^{N \times M}} \gamma \text{ s.t. } \gamma = \frac{1}{K} \sum_{k=1}^{K} \operatorname{Tr} \left(\mathbf{V}^{*} \mathbf{R}_{k} \mathbf{V} \left(\sum_{j=1}^{K} \frac{\mathbf{V}^{*} \mathbf{R}_{j} \mathbf{V}}{1 + \gamma} + \frac{K}{\rho} \mathbf{I}_{M} \right)^{-1} \right)$$

$$= \operatorname{Tr} \left(\mathbf{V}^{*} \mathbf{R}_{\text{tot}} \mathbf{V} \left(\frac{K \mathbf{V}^{*} \mathbf{R}_{i tot} \mathbf{V}}{1 + \gamma} + \frac{K}{\rho} \mathbf{I}_{M} \right)^{-1} \right)$$

$$\text{where } \mathbf{R}_{\text{tot}} = \frac{1}{K} \sum_{k=1}^{K} \mathbf{R}_{k}$$

$$(3) \text{ Optimal solution}$$

$$\mathbf{F}_{\text{RF}} = \mathbf{V} = \text{dominant } M \text{ eigenvectors of } \sum_{k=1}^{K} \mathbf{R}_{k}$$

$$\mathbf{F}_{\text{roof.}} \mu_{1}, \dots, \mu_{M} \text{ eigenvalues of } \mathbf{V}^{*} \mathbf{R}_{\text{tot}} \mathbf{V}$$

$$\lambda_{1}, \dots, \lambda_{N} \text{ eigenvalues of } \mathbf{R}_{\text{tot}} \mathbf{V}$$

$$(3) \text{ optimal solution}$$

$$\mathbf{V} = \frac{1}{K} \sum_{m=1}^{M} \frac{1}{\frac{1}{1 + \gamma} + \frac{1}{\rho \mu_{m}}}$$

$$\leq \frac{1}{K} \sum_{m=1}^{M} \frac{1}{\frac{1}{1 + \gamma} + \frac{1}{\rho \lambda_{m}}}$$

$$(3) \text{ optimal solution}$$

$$\mathbf{V} = \frac{1}{K} \sum_{m=1}^{K} \mathbf{R}_{m} \mathbf{V}$$

$$\mathbf{V} = \frac{1}{K} \sum_{m=1}^{K} \mathbf{R}_{m}$$

$$(4) \text{ optimal solution}$$

$$\mathbf{V} = \frac{1}{K} \sum_{m=1}^{K} \frac{1}{\frac{1}{1 + \gamma}} \mathbf{V} = \frac{1}{\rho \mu_{m}}$$

$$(4) \text{ optimal solution}$$

$$\mathbf{V} = \frac{1}{K} \sum_{m=1}^{K} \frac{1}{\frac{1}{1 + \gamma}} \mathbf{V} = \frac{1}{\rho \mu_{m}}$$

$$(5) \text{ optimal solution}$$

$$\mathbf{V} = \frac{1}{K} \sum_{m=1}^{K} \frac{1}{\frac{1}{1 + \gamma}} \mathbf{V} = \frac{1}{\rho \mu_{m}}$$

$$(5) \text{ optimal solution}$$

$$\mathbf{V} = \frac{1}{K} \sum_{m=1}^{K} \frac{1}{\frac{1}{1 + \gamma}} \mathbf{V} = \frac{1}{\rho \mu_{m}}}$$

$$(5) \text{ optimal solution}$$

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$$(6) \text{ optimal solution}$$

$$(7) \text{ optimal solution}$$

$$(7) \text{ optimal solution}$$

$$(8) \text{ optimal solution}$$

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$$(8) \text{ optimal solu$$



[Proposed 2] Constrained case: Prior work

Prior work ¹⁾

+ Make $\mathbf{F}_{\text{RF,C}}$ (constrained) as similar to $\mathbf{F}_{\text{RF,UC}}$ (unconstrained) as possible

I) Problem formulation

$$\arg\min_{\mathbf{F}_{\mathrm{RF},\mathrm{C}}} \|\mathbf{F}_{\mathrm{RF},\mathrm{UC}} - \mathbf{F}_{\mathrm{RF},\mathrm{C}}\|_{F} \text{ s.t. } \left| \left[\mathbf{F}_{\mathrm{RF},\mathrm{C}}\right]_{i,j} \right| = 1$$

2) Optimal solution

$$\left[\mathbf{F}_{\mathrm{RF},\mathrm{C}}^{\mathrm{opt}}\right]_{i,j} = e^{j \angle \left(\left[\mathbf{F}_{\mathrm{RF},\mathrm{UC}}\right]_{i,j}\right)} \text{ for } i = 1, ..., N, \ j = 1, ..., M$$

Shortcomings of prior work

- When $\mathbf{F}_{\text{RF,UC}}$ is a semi-unitary matrix, the solution loses orthogonality. ($\mathbf{F}_{\text{RF,C}}$ is not a semi-unitary matrix any more.)

1) L. Liang, W. Xu, and X. Dong, "Low-complexity hybrid precoding in massive multiuser MIMO systems," IEEE Wireless Comm. Letters., Dec. 2014.



[Proposed 2] Constrained case: Proposed (1/2)

Proposed

igstarrow Use a compensation matrix that compensates the orthogonality lost in $\mathbf{F}_{\mathrm{RF,C}}$





[Proposed 2] Constrained case: Proposed (2/2)

Step 1. Find an unconstrained analog precoding matrix without phase shifter constraint, $\mathbf{F}_{\rm RF,UC}$

 $\mathbf{F}_{\mathrm{RF,UC}} = M$ dominant eigenvectors of $\sum \mathbf{R}_k$

Step 2. Find a constrained analog precoding matrix with phase shifter constraint, $\mathbf{F}_{\rm RF,C}$

Algorithm 1 Find $\mathbf{F}_{\mathrm{RF,C}}$

Input: $\mathbf{F}_{\mathrm{RF},\mathrm{UC}}$ Initialization: $\mathbf{F}_{(0)} = \measuredangle(\mathbf{F}_{\mathrm{RF},\mathrm{UC}}), n = 0$ **repeat** $n \leftarrow n + 1$ $\mathbf{F}_{(n)} = \measuredangle(\mathbf{F}_{\mathrm{RF},\mathrm{UC}}\mathbf{F}_{\mathrm{RF},\mathrm{UC}}^*\mathbf{F}_{(n-1)})$

until $\|\mathbf{F}_{RF,UC}\mathbf{F}_{RF,UC}^*\mathbf{F}_{(n-1)} - \mathbf{F}_{(n)}\|_F$ converges Output: $\mathbf{F}_{RF,C} = \mathbf{F}_{(n)}$ Step 3. Design a baseband compensation matrix, \mathbf{F}_{CM}

$$\mathbf{F}_{\mathrm{CM}} = \left(\mathbf{F}_{\mathrm{RF,C}}^* \mathbf{F}_{\mathrm{RF,C}}\right)^{-\frac{1}{2}}$$

Step 4. Design a baseband multiuser MIMO precoding matrix, $\mathbf{F}_{\mathrm{MU}}[t]$

Regularized zero-forcing (RZF) with respect to the effective channel $\mathbf{H}_{\text{eff}}^{*}[t] = \mathbf{H}^{*}[t]\mathbf{F}_{\text{RF,C}}\mathbf{F}_{\text{CM}}$

 $\mathbf{F}_{\mathrm{MU}}[t] = \left(\mathbf{H}_{\mathrm{eff}}[t]\mathbf{H}_{\mathrm{eff}}^{*}[t] + \beta \mathbf{I}\right)^{-1}\mathbf{H}_{\mathrm{eff}}[t]$

Step 5. Design a final analog precoding matrix \mathbf{F}_{RF} and a baseband precoding matrix $\mathbf{F}_{\text{BB}}[t]$

$$\mathbf{F}_{\rm RF} = \mathbf{F}_{\rm RF,C}, \quad \mathbf{F}_{\rm BB}[t] = \mathbf{F}_{\rm CM} \mathbf{F}_{\rm MU}[t]$$



Simulation results



- N = 64, M = K
- L = 5, $\sigma_{AS} = 10^{\circ}$, SNR = 10 dB
- w/o phase shifter constraint

Sum spectral efficiency vs. K



- $N = 64, M = \{16, 32, 48\}$
- L = 5, $\sigma_{AS} = 10^{\circ}$, SNR = 10 dB
- w/ & w/o phase shifter constraint



Conclusions

Our proposed hybrid precoding technique for massive MIMO systems

Uses only long-term channel statistics for the analog precoder design

Mitigates the loss caused by using phase shifters



Thank you !



