

# Main Presentation

## Hybrid Codebook Construction and Sum Rate Maximization in mmWave Multiuser Systems

Vaibhav Baranwal

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Department of Electrical Communication Engineering  
Indian Institute of Science

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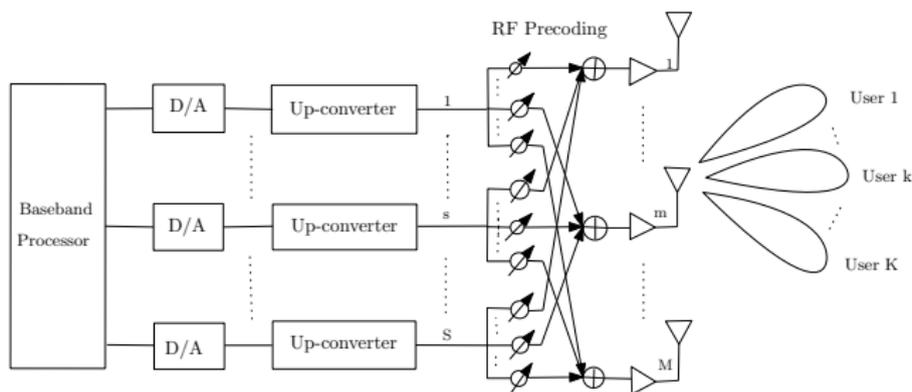
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- To reduce the losses we use beamforming techniques to generate beamforming vectors
- We have simulated a hybrid precoding method from the literature to generate the RF precoder and baseband precoder to select the beamformers[1]

# System Model

Consider a multiple input single output (MISO) system,



The system consists of a baseband processor, digital-to-analog converter (DAC), upconverter,  $S$  RF chains,  $M$  transmit antennas, and  $K$  number of users.

The system model is expressed as

$$\mathbf{y} = \mathbf{H}\mathbf{F}\mathbf{G}\mathbf{s} + \mathbf{n}$$

Here,

**H**: channel matrix of the system of size  $K \times M$ ,

**F**: RF precoder of size  $M \times S$ ,

**G**: baseband precoder of size  $S \times K$ ,

**s**: transmitted vector of size  $K \times 1$ ,

**n**: noise vector of size  $K \times 1$ ,

**y**: received vector of size  $K \times 1$ .

Here,  $M \geq S \geq K$

# Channel Model

Here,  $\mathbf{s}^T = [s_1, \dots, s_K]$  with  $s_k \sim \mathcal{CN}(0, 1)$ ,  $\forall k \in \mathcal{K} \triangleq \{1, 2, \dots, K\}$

$\mathbf{H}^H = [\mathbf{h}_1, \dots, \mathbf{h}_K]$ . Here,  $\mathbf{h}_k$  is given by

$$\mathbf{h}_k = \sqrt{\frac{M}{N_{cl} N_{ray}}} \sum_{m_p=1}^{N_{cl}} \sum_{n_p=1}^{N_{ray}} \alpha_{m_p, n_p} \mathbf{a}(\phi_{m_p, n_p})$$

$$\text{where, } \mathbf{a}_{ULA}(\phi) = \sqrt{\frac{1}{M}} \left[ 1, e^{j\frac{2\pi}{\lambda} d \sin(\phi)}, \dots, e^{j(M-1)\frac{2\pi}{\lambda} d \sin(\phi)} \right]$$

$\lambda$  is the signal wavelength

$d$  is the antenna spacing

$\phi$  is the angle of departure (AoD)

$\mathbf{n} \sim \mathcal{CN}(\mathbf{0}, \sigma^2 \mathbf{I}_K)$  is the AWGN noise vector.

$\alpha_{m_p, n_p} \sim \mathcal{CN}(0, \sigma_{m_p}^2)$

# Sum Rate Expression

Problem of maximizing system sum rate is given as

$$\begin{aligned} & \max_{\underline{\mathbf{F}}, \underline{\mathbf{G}}} \sum_{k=1}^K \underline{R}_k \\ \text{s.t. } & \underline{R}_k = \log(1 + \underline{\text{SINR}}_k) \geq \gamma_k, \forall k \in \mathcal{K} \\ & \underline{\mathbf{F}} \in \mathcal{F}_{RF}, \|\underline{\mathbf{F}}\underline{\mathbf{G}}\|_{\mathcal{F}}^2 \leq P \\ & \underline{\text{SINR}}_k = \frac{|\mathbf{h}_k^H \underline{\mathbf{F}} \underline{\mathbf{g}}_k|^2}{\sum_{l=1, l \neq k}^K |\mathbf{h}_k^H \underline{\mathbf{F}} \underline{\mathbf{g}}_l|^2 + \sigma^2} \end{aligned}$$

Some examples of the RF codebooks are

- $q$ -bit quantizer Codebook

$$\mathbf{F}(m, n) = \frac{1}{\sqrt{M}} e^{j \frac{\pi(4(m-1)(n-1)-2N)}{2^{q+1}}}, \forall m \in \mathcal{M}, \forall n \in \mathcal{N}$$

- IEEE 802.15.3c Codebook

$$\mathbf{F}(m, n) = \frac{1}{\sqrt{M}} e^{j \frac{\pi}{2} \text{floor} \left( \frac{4(m-1) \left( \text{mod} \left( (n-1) + \frac{N}{4}, N \right) \right)}{N} \right)}, \forall m \in \mathcal{M}, \forall n \in \mathcal{N}$$

- DFT Codebook

$$\mathbf{F}(m, n) = \frac{1}{\sqrt{M}} e^{-j \frac{2\pi(m-1)(n-1)}{M}}, \forall m \in \mathcal{M}, \forall n \in \mathcal{N}$$

Here,  $\mathcal{M} \in \{1, \dots, M\}$ ,  $\mathcal{N} \in \{1, \dots, N\}$

- DFT-based Multilevel Codebook [2]

$$\begin{aligned} \mathcal{F}_m &= \left\{ f_1^{(m)}, f_2^{(m)}, \dots, f_{M/N}^{(m)} \right\} \\ &= \left\{ \frac{1}{\sqrt{N}} \sum_{p=1}^N \mathbf{u}_t(p) e^{j\omega_m p}, \frac{1}{\sqrt{N}} \sum_{p=N+1}^{2N} \mathbf{u}_t(p) e^{j\omega_m p}, \dots, \right. \\ &\quad \left. \frac{1}{\sqrt{N}} \sum_{p=M-N+1}^M \mathbf{u}_t(p) e^{j\omega_m p} \right\} \end{aligned}$$

$$\mathbf{u}_t(n) = \frac{1}{\sqrt{M}} \left[ 1, e^{-j\frac{2\pi}{M}\omega_m(n-\frac{M+1}{2})}, e^{-j\frac{2\pi}{M}\omega_m 2(n-\frac{M+1}{2})}, \dots, e^{-j\frac{2\pi}{M}\omega_m(M-1)(n-\frac{M+1}{2})} \right]$$

$\omega_m \in \left[ -\frac{\pi}{M}, \pi \left( 1 - \frac{1}{M} \right) \right]$  and is selected by minimizing  $\text{var} \left( \left| \mathbf{u}_t^H(n) \mathbf{f}_k^{(m)} \right| \right)$ .

# Beam Sweep Procedure

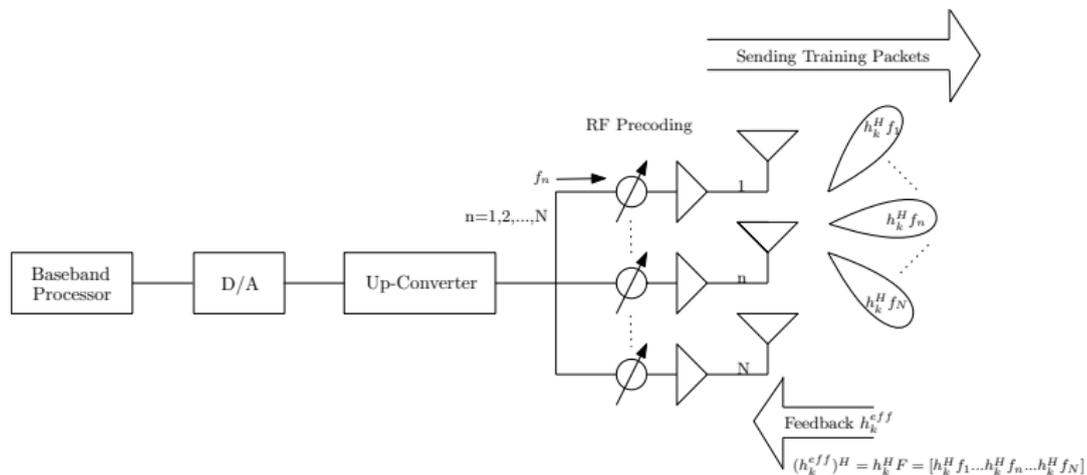


Figure: Beam-sweep Procedure

# Virtual Communication System for CSIT

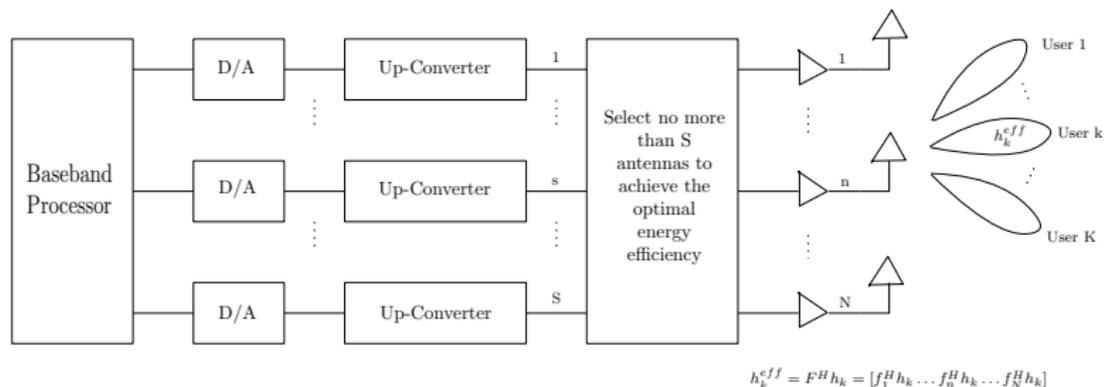


Figure: Virtual Multiuser Communication System

# Baseband Precoder Design

Sum rate maximization problem is expressed as

$$\begin{aligned} \max_{\mathbf{G}} \quad & \sum_{k=1}^K R_k, \\ \text{s.t.} \quad & R_k = \log(1 + \text{SINR}_k), \\ & \sum_{k=1}^K \|\mathbf{F}\mathbf{g}_k\|_2^2 \leq P, \|\ddot{\mathbf{g}}\|_0 \leq S, \end{aligned}$$

where,  $\ddot{\mathbf{g}} = [\|\tilde{\mathbf{g}}_1\|_2, \dots, \|\tilde{\mathbf{g}}_N\|_2]^T$

$$\text{SINR}_k = \frac{|\mathbf{h}_k^H \mathbf{F}\mathbf{g}_k|^2}{\sum_{l=1, l \neq k}^K |\mathbf{h}_k^H \mathbf{F}\mathbf{g}_l|^2 + \sigma^2}$$

We will introduce some variables  $a_k, b_k, \forall k \in \mathcal{K}$ , So, the sum rate maximization problem can be rewritten as

$$\begin{aligned} & \min_{\{\mathbf{g}_k, a_k, b_k\}} - \sum_{k=1}^K b_k, \\ \text{s.t. } & 1 + a_k \geq e^{b_k}, \forall k \in \mathcal{K}, \\ & \text{SINR}_k \geq a_k, \text{ SINR}_k \geq \bar{\gamma}_k, \forall k \in \mathcal{K}, \\ & \sum_{k=1}^K \|\mathbf{F}\mathbf{g}_k\|_2^2 \leq P, \|\mathbf{g}\|_0 \leq S, \end{aligned}$$

where  $\bar{\gamma}_k = e^{\gamma_k} - 1$

# Rearranging the problem

To overcome nonconvex difficulties

$$\begin{aligned} \min_{\{\mathbf{g}_k, a_k, b_k\}} & -\sum_{k=1}^K b_k + \lambda \|\mathbf{g}\|_0, \\ \text{s.t. } & 1 + a_k \geq e^{b_k}, \forall k \in \mathcal{K}, \sum_{k=1}^K \|\mathbf{F}\mathbf{g}_k\|_2^2 \leq P, \\ & \text{SINR}_k \geq a_k, \text{SINR}_k \geq \bar{\gamma}_k, \forall k \in \mathcal{K}, \end{aligned}$$

Here,  $\lambda$  control the sparsity of solution, i.e., the larger  $\lambda$  the solution is more sparse.

# Approximate Problem

Problem can be further approximated as

$$\begin{aligned} \min_{\{\mathbf{g}_k, a_k, b_k\}} & - \sum_{k=1}^K b_k + \lambda \|\mathbf{G}\|_{1,\infty}^2, \\ \text{s.t. } & 1 + a_k \geq e^{b_k}, \forall k \in \mathcal{K}, \sum_{k=1}^K \|\mathbf{F}\mathbf{g}_k\|_2^2 \leq P, \\ & \text{SINR}_k \geq a_k, \text{ SINR}_k \geq \bar{\gamma}_k, \forall k \in \mathcal{K}, \end{aligned}$$

where  $\|\mathbf{G}\|_{1,\infty} = \sum_{n=1}^N \max_k |\mathbf{g}_k(n)|$  is the  $l_{1,\infty}$ -norm of matrix  $\mathbf{G}$ .

Here,  $\|\mathbf{G}\|_{1,\infty}^2$  can be rewritten as

$$\begin{aligned}\|\mathbf{G}\|_{1,\infty}^2 &= \left( \sum_{n=1}^N \max_k |\mathbf{g}_k(n)| \right)^2, \\ &= \sum_{n=1}^N \sum_{m=1}^N \left( \left( \max_k |\mathbf{g}_k(n)| \right) \left( \max_k |\mathbf{g}_k(m)| \right) \right), \\ &= \sum_{n=1}^N \sum_{m=1}^N \max_{i,j \in \{1,\dots,K\}} |\mathbf{X}_{i,j}(n, m)|,\end{aligned}$$

where  $\mathbf{X}_{i,j} = \mathbf{g}_i \mathbf{g}_j^H$ ,  $\forall i, j$ . So,  $\mathbf{X}_{i,i} = \mathbf{g}_i \mathbf{g}_i^H$ , therefore  $\mathbf{X}_{i,i} \succeq 0$  and  $\text{rank}(\mathbf{X}_{i,i}) = 1$ ,  $\forall i$ .

# Approximate Problem

Problem can be further approximated as

$$\begin{aligned} \min_{\{\mathbf{X}_{i,j}, a_k, b_k\}} & - \sum_{k=1}^K b_k + \lambda \|\mathbf{G}\|_{1,\infty}^2, \\ \text{s.t. } & 1 + a_k \geq e^{b_k}, \forall k \in \mathcal{K}, \sum_{k=1}^K \text{tr}(\tilde{\mathbf{F}}\mathbf{X}_{k,k}) \leq P, \\ & \text{SINR}_k \geq a_k, \text{SINR}_k \geq \bar{\gamma}_k, \forall k \in \mathcal{K}, \mathbf{X}_{k,k} \succeq \mathbf{0}, \forall k \in \mathcal{K}, \\ & \text{rank}(\mathbf{X}_{i,j}) = 1, \forall i, j, \end{aligned}$$

where  $\tilde{\mathbf{F}} = \mathbf{F}^H \mathbf{F}$ , and

$$\text{SINR}_k = \frac{\text{tr}(\mathbf{H}_k \mathbf{X}_{k,k})}{\sum_{l=1, l \neq k}^K \text{tr}(\mathbf{H}_k \mathbf{X}_{l,l}) + \sigma^2}$$

where  $\mathbf{H}_k = \mathbf{F}^H \mathbf{h}_k \mathbf{h}_k^H \mathbf{F}$ ,  $\forall k \in \mathcal{K}$ .

# Further Approximation

Let  $\mathbf{X}_k = \mathbf{X}_{k,k}$ ,  $\forall k \in \mathcal{K}$  and  $\mathbf{Z}(n, m) = \max_{k \in \mathcal{K}} |\mathbf{X}_k(n, m)|$ ,  $\forall k \in \mathcal{K}$ .

We will drop the nonconvex constraints  $\text{rank}(\mathbf{X}_k) = 1$ .

$$\begin{aligned} & \min_{\{\mathbf{X}_k, a_k, b_k\}, \mathbf{Z}} - \sum_{k=1}^K b_k + \lambda \text{tr}(\mathbf{1}_{N \times N} \mathbf{Z}) \\ \text{s.t. } & 1 + a_k \geq e^{b_k}, \forall k \in \mathcal{K}, \sum_{k=1}^K \text{tr}(\tilde{\mathbf{F}} \mathbf{X}_k) \leq P, \\ & \text{SINR}_k \geq a_k, \text{SINR}_k \geq \bar{\gamma}_k, \forall k \in \mathcal{K}, \\ & \mathbf{X}_k \succeq \mathbf{0}, \mathbf{Z} \geq |\mathbf{X}_k|, \forall k \in \mathcal{K}, \end{aligned}$$

We will introduce some new variables  $\psi_k, \phi_k, \forall k \in \mathcal{K}$ . The problem can be rewritten as

$$\min_{\{\mathbf{X}_k, a_k, b_k, \psi_k, \phi_k\}, \mathbf{Z}} - \sum_{k=1}^K b_k + \lambda \text{tr}(\mathbf{1}_{N \times N} \mathbf{Z}),$$

$$\text{s.t. } \psi_k^2 \leq \text{tr}(\mathbf{H}_k \mathbf{X}_k), \mathbf{X}_k \succeq \mathbf{0}, 1 + a_k \geq e^{b_k}, \forall k \in \mathcal{K}$$

$$\sum_{l=1, l \neq k}^K \text{tr}(\mathbf{H}_k \mathbf{X}_{l,l}) + \sigma^2 \leq \phi_k, \forall k \in \mathcal{K}, \sum_{k=1}^K \text{tr}(\tilde{\mathbf{F}} \mathbf{X}_k) \leq P,$$

$$\sum_{l=1, l \neq k}^K \bar{\gamma}_k \text{tr}(\mathbf{H}_k \mathbf{X}_{l,l}) + \bar{\gamma}_k \sigma^2 \leq \text{tr}(\mathbf{H}_k \mathbf{X}_k), \frac{\psi_k^2}{\phi_k} \geq a_k, \forall k \in \mathcal{K},$$

$$\begin{bmatrix} \mathbf{Z}(n, m) - \Re(\mathbf{X}_k(n, m)) & \Im(\mathbf{X}_k(n, m)) \\ \Im(\mathbf{X}_k(n, m)) & \mathbf{Z}(n, m) + \Re(\mathbf{X}_k(n, m)) \end{bmatrix} \succeq \mathbf{0}, \forall k \in \mathcal{K}, m, n.$$

# Final Convex Problem

The constraint  $\frac{\psi_k^2}{\phi_k} \geq a_k$  is still nonconvex. So, the constraint can be approximated as

$$\frac{\psi_k^2}{\phi_k} \geq \Phi_k^{(l)}(\psi_k, \phi_k) \triangleq 2 \frac{\psi_k^{(l)}}{\phi_k^{(l)}} \psi_k - \left( \frac{\psi_k^{(l)}}{\phi_k^{(l)}} \right)^2 \phi_k, \forall k \in \mathcal{K},$$

Now, we will solve the series of convex optimization problem. Here,  $l$  denotes the  $l$ th iteration.  $(\psi_k^{(l)}, \phi_k^{(l)}) \leftarrow (\psi_k, \phi_k)$  at the  $l$ th iteration.  $\Phi_k^{(l)}(\psi_k, \phi_k)$  is determined at the  $l$ th iteration.

So, the approximate value of convex problem at the  $(l + 1)$ th iteration can be

$$\min_{\{\mathbf{X}_k, a_k, b_k, \psi_k, \phi_k\} \mathbf{Z}} - \sum_{k=1}^K b_k + \lambda \text{tr}(\mathbf{1}_{N \times N} \mathbf{Z}),$$

$$\text{s.t. } \psi_k^2 \leq \text{tr}(\mathbf{H}_k \mathbf{X}_k), \mathbf{X}_k \succeq \mathbf{0}, 1 + a_k \geq e^{b_k}, \forall k \in \mathcal{K},$$

$$\sum_{l=1, l \neq k}^K \text{tr}(\mathbf{H}_k \mathbf{X}_l) + \sigma^2 \leq \phi_k, \forall k \in \mathcal{K}, \sum_{k=1}^K \text{tr}(\tilde{\mathbf{F}} \mathbf{X}_k) \leq P,$$

$$\sum_{l=1, l \neq k}^K \bar{\gamma}_k \text{tr}(\mathbf{H}_k \mathbf{X}_l) + \bar{\gamma}_k \sigma^2 \leq \text{tr}(\mathbf{H}_k \mathbf{X}_k),$$

$$\Phi_k^{(l)}(\psi_k, \phi_k) \geq a_k, \forall k \in \mathcal{K},$$

$$\begin{bmatrix} \mathbf{Z}(n, m) - \Re(\mathbf{X}_k(n, m)) & \Im(\mathbf{X}_k(n, m)) \\ \Im(\mathbf{X}_k(n, m)) & \mathbf{Z}(n, m) + \Re(\mathbf{X}_k(n, m)) \end{bmatrix} \succeq \mathbf{0}, \forall k \in \mathcal{K}, m, n.$$

## Algorithm for optimal solution

Fix a value of  $\lambda$ . Let the value of our objective function is  $\tau$  and  $\tau^{(l)}$  is the value of  $\tau$  at the  $l$ th iteration.

- 1: Let  $l = 0$ , take some initial points as  $\Gamma^{(l)}$  and get  $\tau^{(l)}$ .
- 2: Solve the convex problem with  $\Gamma^{(l)}$ , and obtain new values of the  $\Gamma$  and  $\tau$ .
- 3: If  $|\tau - \tau^{(l)}| \leq \zeta$ , then  $\tau, \Gamma$  will be our optimal solution, otherwise  $\Gamma^{(l)} \leftarrow \Gamma, \tau^{(l)} \leftarrow \tau$  and go to step 2.

# Selection of $\lambda$

Let  $L^\lambda$  be the number of nonzero diagonal entries in  $\mathbf{Z}$ .

## Algorithm for choosing $\lambda$

- 1: Generate initial points  $\lambda_L, \lambda_U$  and compute  $\tilde{\tau}^T = \sum_{k=1}^K b_k$  and denote  $\Xi^T$  as the temporary solution of the convex problem. Let  $flag = 1$
- 2: **while** flag **do**
- 3: Let  $\lambda = \frac{\lambda_L + \lambda_U}{2}$ .
- 4: Solve the convex problem with  $\lambda$ , then obtain the solution of it after iteration and  $\tilde{\tau}^\lambda$
- 5: If  $L^\lambda > S$ , let  $\lambda_L = \lambda$ , otherwise, let  $\lambda_U = \lambda$ .
- 6: If  $|\tilde{\tau}^\lambda - \tilde{\tau}^T| \leq \zeta$  and  $L^\lambda \leq S$ , then let  $flag=0$  and output the solution of convex problem. Otherwise,  $\Xi^T \leftarrow \Xi^\lambda, \tilde{\tau}^T \leftarrow \tilde{\tau}^\lambda$
- 7: **end while**

# Refining Solution

$\hat{\mathbf{F}}$  is obtained by choosing  $L^\lambda$  codewords from RF codebook.

$\bar{\mathbf{h}}_k = \hat{\mathbf{F}}^H \mathbf{h}_k$  is our effective channel.

Now we will refine our solution

$$\begin{aligned} & \max_{\{\bar{\mathbf{g}}_k\}} \sum_{k=1}^K \bar{R}_k, \\ & \text{s.t. } \overline{\text{SINR}}_k \geq \bar{\gamma}_k, \forall k \in \mathcal{K}, \\ & \sum_{k=1}^K \|\hat{\mathbf{F}} \bar{\mathbf{g}}_k\|_2^2 \leq P, \end{aligned}$$

where  $\bar{R}_k = \log(1 + \overline{\text{SINR}}_k)$ , and  $\overline{\text{SINR}}_k$  is given by

$$\overline{\text{SINR}}_k \triangleq \frac{\|\bar{\mathbf{h}}_k^H \bar{\mathbf{g}}_k\|_2^2}{\sum_{l=1, l \neq k}^K \|\bar{\mathbf{h}}_k^H \bar{\mathbf{g}}_l\|_2^2 + \sigma^2}.$$

# Rearranging the Problem

We introduce variable  $\bar{a}_k, \bar{b}_k, \bar{\phi}_k$ , the problem can be reformulated as

$$\max_{\{\bar{\mathbf{g}}_k, \bar{a}_k, \bar{b}_k, \bar{\phi}_k\}} \sum_{k=1}^K \bar{b}_k,$$

$$\text{s.t. } 1 + \bar{a}_k \geq e^{\bar{b}_k}, \forall k \in \mathcal{K}, \sum_{k=1}^K \|\hat{F} \bar{\mathbf{g}}_k\|_2^2 \leq P$$

$$\frac{\|\bar{\mathbf{h}}_k^H \bar{\mathbf{g}}_k\|_2^2}{\bar{\phi}_k} \geq \bar{\gamma}_k, \forall \frac{\|\bar{\mathbf{h}}_k^H \bar{\mathbf{g}}_k\|_2^2}{\bar{\phi}_k} \geq \bar{a}_k, \forall k \in \mathcal{K},$$

$$\sum_{l=1, l \neq k}^K \|\bar{\mathbf{h}}_k^H \bar{\mathbf{g}}_l\|_2^2 + \sigma^2 \leq \bar{\phi}_k, \forall k \in \mathcal{K},$$

# Final Sum Rate Maximization Problem

As the problem is nonconvex, we can approximate the constraint by

$$\frac{\|\bar{h}_k^H \bar{g}_k\|_2^2}{\bar{\phi}_k} \geq \bar{\Phi}_k^{(l)}(\bar{g}_k, \bar{\phi}_k) \triangleq \frac{2\Re\left(\left(\bar{g}_k^{(l)}\right)^H \bar{h}_k \bar{h}_k^H \bar{g}_k\right)}{\bar{\phi}_k^{(l)}} - \left(\frac{\|\bar{h}_k^H \bar{g}_k^{(l)}\|_2}{\bar{\phi}_k^{(l)}}\right)^2 \bar{\phi}_k,$$
$$\forall k \in \mathcal{K},$$

Here  $l$  denotes the  $l$ th iteration. We can write

$$\bar{\Phi}_k^{(l)}(\bar{g}_k, \bar{\phi}_k) \geq \bar{\gamma}_k, \quad \bar{\Phi}_k^{(l)}(\bar{g}_k, \bar{\phi}_k) \geq \bar{a}_k, \quad \forall k \in \mathcal{K}.$$

We solve the following convex problem to obtain the optimal solution

$$\begin{aligned} & \max_{\{\bar{\mathbf{g}}_k, \bar{a}_k, \bar{b}_k, \bar{\phi}_k\}} \sum_{k=1}^K \bar{b}_k, \\ \text{s.t. } & 1 + \bar{a}_k \geq e^{\bar{b}_k}, \forall k \in \mathcal{K}, \sum_{k=1}^K \|\hat{F} \bar{\mathbf{g}}_k\|_2^2 \leq P \\ & \sum_{l=1, l \neq k}^K \|\bar{\mathbf{h}}_k^H \bar{\mathbf{g}}_l\|_2^2 + \sigma^2 \leq \bar{\phi}_k, \forall k \in \mathcal{K}, \\ & \bar{\Phi}_k^{(l)}(\bar{\mathbf{g}}_k, \bar{\phi}_k) \geq \bar{\gamma}_k, \bar{\Phi}_k^{(l)}(\bar{\mathbf{g}}_k, \bar{\phi}_k) \geq \bar{a}_k, \forall k \in \mathcal{K}. \end{aligned}$$

## Algorithm for optimal solution

Let the value of our objective function is  $\bar{\tau}$  and  $\bar{\tau}^{(l)}$  is the value of  $\bar{\tau}$  at the  $l$ th iteration.

- 1: Let  $l = 0$ , take some initial points as  $\bar{\Gamma}^{(l)}$  and get  $\bar{\tau}^{(l)}$ .
- 2: Solve the convex problem with  $\bar{\Gamma}^{(l)}$ , and obtain new values of the  $\bar{\Gamma}$  and  $\bar{\tau}$ .
- 3: If  $|\bar{\tau} - \bar{\tau}^{(l)}| \leq \zeta$ , then  $\bar{\tau}, \bar{\Gamma}$  will be our optimal solution, otherwise  $\bar{\Gamma}^{(l)} \leftarrow \bar{\Gamma}, \bar{\tau}^{(l)} \leftarrow \bar{\tau}$  and go to step 2.

# Simulation Result

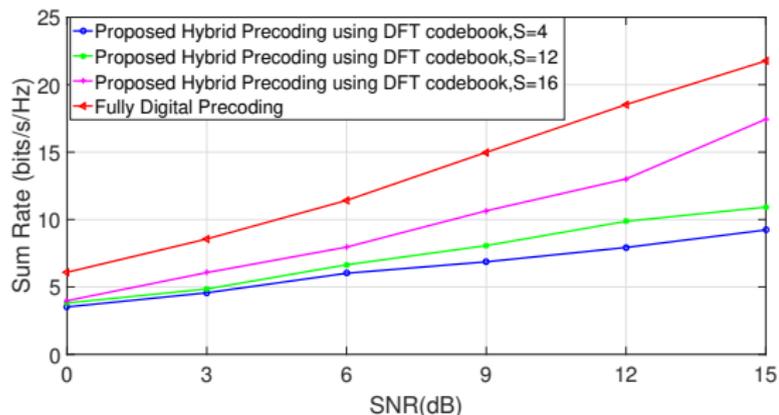


Figure: Sum Rate vs SNR,  $M = N = 16$ ,  $K = 4$

- ULA of transmit antennas  
 $d = \lambda/2$
- $N_{cl} = 6$ ,  $N_{ray} = 8$   
AoD  $\sim$  Laplacian, mean of  $\phi_{m_p} \sim \mathcal{U}[-\pi, \pi)$ ,  $\sigma_\phi = 7.5^\circ$
- $\zeta = 10^{-3}$
- $\lambda_L = 0$  &  $\lambda_U = 100$ .

# Simulation Result

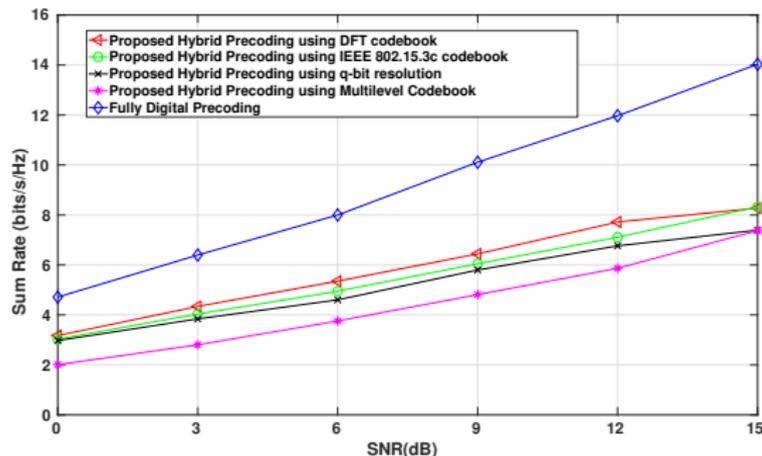


Figure: Sum Rate vs SNR,  $M = N = 16$ ,  $S = 4$ ,  $K = 2$

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AoD  $\sim$  Laplacian,  
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 $\sigma_\phi = 7.5^\circ$
- $\zeta = 10^{-3}$
- $\lambda_L = 0$  &  
 $\lambda_U = 100$ .

- In the future work we will optimize the codebook to get the best RF codebook among different codebooks.
- Also, we will check the robustness of the system when there are analog errors in RF precoder.

-  S. He, J. Wang, Y. Huang, and W. Hong, “Codebook based hybrid precoding for millimeter wave multiuser systems,” *IEEE Trans. Signal Process.*, vol. 65, no. 20, pp. 5289–5304, Oct. 15 2017.
-  S. Noh, M. D. Zoltowski, and D. J. Love, “Multi-resolution codebook and adaptive beamforming sequence design for millimeter wave beam alignment,” *IEEE Trans. Wireless Commun.*, vol. 16, no. 9, pp. 5689–1683, Sep. 2017.

Thank You