

教育部下世代尖端無線技術聯盟計畫

5G大型陣列天線基頻模組課程

Part I: 理論部分

大綱

- 理論一: 5G Massive MIMO Baseband Processing
(**Channel, ABF, Precoding, Hybrid-BF**) (頁數: pp. 3-76)
 - 1. Channel Model for mmWave Massive MIMO
 - 2. Digital Precoding
 - 3. Analog Beamforming (ABF)
 - 4. Hybrid Precoding

- 理論二: 5G Massive MIMO Baseband Processing
(**DBF, DOA, MU SYNC**) (頁數: pp. 77-158)
 - 1. Digital Beamforming (DBF)
 - 2. DOA Estimation
 - 3. Multiuser Synchronizations (MU SYNC)

理論一

5G Massive MIMO Baseband Processing (Channel, ABF, Precoding, Hybrid-BF)

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Outline

- 1. Introduction
- 2. Channel Model for mmWave Massive MIMO
- 3. Digital Precoding
 - Single User DP
 - Multiuser DP
- 4. Analog Beamforming (ABF)
 - Beam Steering
 - Beam Training (Beam Alignment)
- 5. Hybrid Precoding
 - Single User HP
 - Multiuser HP
- 6. Conclusions

1. Introduction

- The fundamentals of **precoding (beamforming)** are the same regardless of carrier frequency.
 - **Signal processing** in **mmWave massive MIMO** systems needs to be subject to a practical constraints.
- For the traditional **analog beamforming** with **small RF chains**:
 - Analog circuit with **phase shifter (PS)** network
 - Constant amplitude constraints
 - Suffering the performance loss (**Interference Problems**)
- For the **digital precoding** with large antennas and RF chains:
 - High cost and energy consumption (**large number of RF chains**)

Introduction

- Controlling both **phase and amplitude** to cancel interferences and achieve the optimal performance.
- For the **hybrid** analog and digital precoding:
 - **Small RF chains** with **small** digital **precoder** to cancel interferences.
 - **Large analog beamformer** with large PS to increase the antenna array gain.
 - Significantly reducing the RF chains **without obvious performance loss**.

2. Channel Model for mmWave Massive MIMO

- **High free-space path loss** is a characteristic of mmWave propagation.
 - Limited spatial selectivity or scattering.
- **Large tightly packed antenna** arrays are characteristics of mmWave transceivers.
 - **High** levels of **antenna correlation**.
- Characteristics in mmWave channels.
 - **Sparse scattering channel**
 - Adopting a narrowband clustered channel representation, based on the **extended Saleh-Valenzuela model**.

Channel Model for mmWave Massive MIMO

- Using the **clustered channel model**, the discrete-time channel matrix \mathbf{H} can be expressed by a sum of the L propagation paths, i.e.,

$$\mathbf{H} = \sqrt{\frac{N_t N_r}{L}} \sum_{l=1}^L \alpha_l \Lambda_r(\phi_l^r, \theta_l^r) \Lambda_t(\phi_l^t, \theta_l^t) \mathbf{a}_r(\phi_l^r, \theta_l^r) \mathbf{a}_t^H(\phi_l^t, \theta_l^t)$$

where

α_l : complex gain of the l_{th} path

$\phi_l^r(\theta_l^r)$: AZ/EL angles of AoAs

$\phi_l^t(\theta_l^t)$: AZ/EL angles of AoDs

$\Lambda_r(\phi_l^r, \theta_l^r)$: RX antenna element gain at l_{th} AoA, $\Lambda_r = 1$ for simplicity

$\Lambda_t(\phi_l^t, \theta_l^t)$: TX antenna element gain at l_{th} AoD, $\Lambda_t = 1$ for simplicity

Channel Model for mmWave Massive MIMO

- For the uniform linear array (ULA) with N elements, the normalized array response vector:

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

- For the uniform planar array (UPA) with W_1 and W_2 elements ($W_1 W_2 = N$), the normalized array response vector:

$$\mathbf{a}_{\text{UPA}}(\phi, \theta) = \frac{1}{\sqrt{N}} \left[1, \dots, e^{j\frac{2\pi}{\lambda}d(x\sin(\phi)\sin(\theta) + y\cos(\theta))}, \dots, e^{j\frac{2\pi}{\lambda}d((W_1-1)\sin(\phi)\sin(\theta) + (W_2-1)\cos(\theta))} \right]^T$$

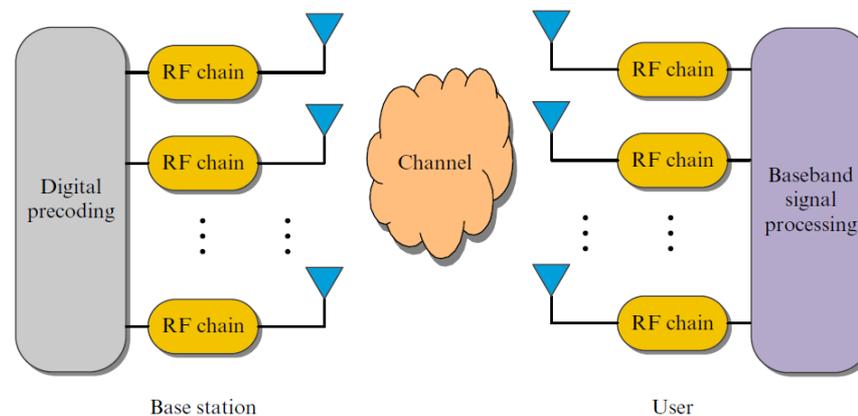
where $0 \leq x \leq W_1 - 1$ (horizontal) and $0 \leq y \leq W_2 - 1$ (vertical)

3. Digital Precoding (DP)

- ❑ DP can control “**phases and amplitudes**” of original signals to cancel interferences.
- ❑ Two categories of DP:
 - **Linear precoding**: TX signals with the linear combination of the original signals.
 - **Nonlinear precoding**: TX signals with the nonlinear processing.
- ❑ Two system used by DP techniques:
 - **Single-user** precoding system: Matched Filter (**MF**) and Zero-Forcing (**ZF**) precoding.
 - **Multiuser** precoding system: Block diagonalization (**BD**) precoding.

3.1 Single-User Digital Precoding

- Architecture of DP for single-user mmWave massive MIMO system:



Architecture of digital precoding for single-user mmWave massive MIMO system.

- Consider N_t TX antennas, N_r RX antennas, N_r data streams ($N_r < N_t$), the TX signal with $N_t \times N_r$ DP \mathbf{D} matrix:

$$\mathbf{x} = \mathbf{D}\mathbf{s}$$

where

\mathbf{s} : $N_r \times 1$ original signal vector with $E(\mathbf{s}\mathbf{s}^H) = \frac{1}{N_r} \mathbf{I}_{N_r}$

\mathbf{D} : precoding matrix to meet total TX power constraint, $\|\mathbf{D}_F\|^2 = \text{tr}(\mathbf{D}\mathbf{D}^H) = N_r$

Single-User Digital Precoding

- Under the narrowband system, the RX signal vector $\mathbf{y}(N_r \times 1)$:

$$\mathbf{y} = \sqrt{\rho} \mathbf{H} \mathbf{D} \mathbf{s} + \mathbf{n},$$

where

\mathbf{H} : $N_r \times N_t$ channel matrix with normalized power $E(\|\mathbf{H}\|_F^2) = N_t N_r$

ρ : The averaged RX power

\mathbf{n} : AWGN noise vector, i.i.d. noise element $CN(0, \sigma_n^2)$

\mathbf{H} : Assume H known at BS to enable precoding

Single-User Digital Precoding

- The simplest linear digital precoding: **MF precoding**

$$\mathbf{D} = \sqrt{\frac{N_r}{\text{tr}(\mathbf{F}\mathbf{F}^H)}}\mathbf{F},$$

$$\mathbf{F} = \mathbf{H}^H.$$

- MF can maximize the SNR at user side.
 - MF involves severe interferences among different data streams.
- The well-known linear digital precoding: **ZF precoding**

$$\mathbf{D} = \sqrt{\frac{N_r}{\text{tr}(\mathbf{F}\mathbf{F}^H)}}\mathbf{F},$$

$$\mathbf{F} = \mathbf{H}^H (\mathbf{H}\mathbf{H}^H)^{-1}$$

Single-User Digital Precoding

- ZF can entirely eliminate the interferences among different data streams.
- \mathbf{D} required to satisfy the total TX power constraint. ZF precoding may **enhance the power of noise** and lead performance loss.
- The Wiener filter (WF) precoding: **WF precoding** (or **MMSE precoding**)

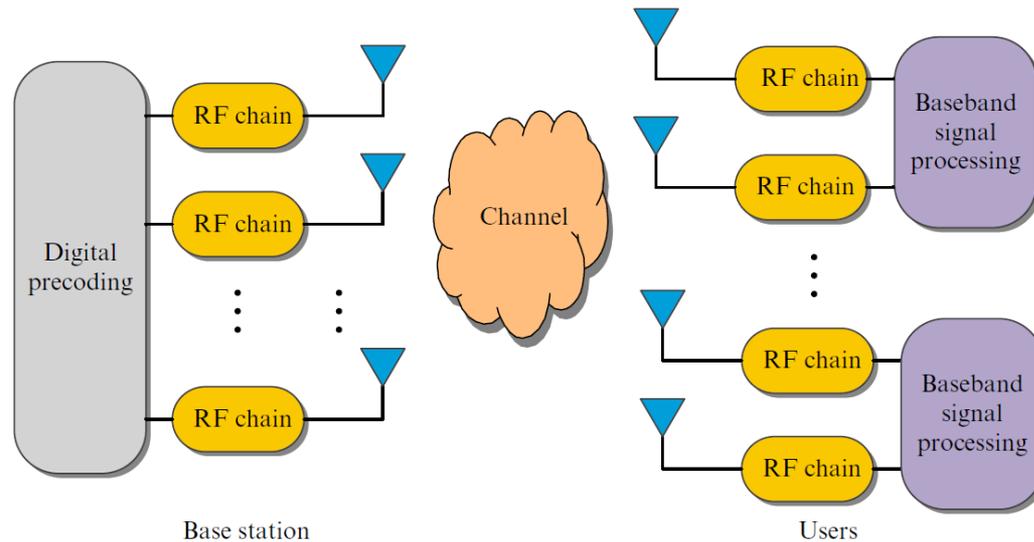
$$\mathbf{D} = \sqrt{\frac{N_r}{\text{tr}(\mathbf{F}\mathbf{F}^H)}} \mathbf{F},$$

$$\mathbf{F} = \mathbf{H}^H \left(\mathbf{H}\mathbf{H}^H + \frac{\sigma_n^2 N_r}{\rho} \mathbf{I} \right)^{-1}$$

.. WF precoding can make a better trade-off between the RX SNR and interferences.

3.2 Multiuser Digital Precoding

- Architecture of digital precoding for multiuser mmWave massive MIMO Systems.



Architecture of digital precoding for multiuser mmWave massive MIMO systems.

- Consider BS with N_{BS} antennas and RF chains.
- Communicate with U MSs, each MS with N_{MS} antennas.
- Total data streams is $N_{MS}U$ ($N_{MS}U \leq N_{BS}$)

Multiuser Digital Precoding

- For downlink communication, BS employs U digital precoders, i.e., $\mathbf{D}=[\mathbf{D}_1 \mathbf{D}_2 \dots \mathbf{D}_U]$, where \mathbf{D}_u is the $N_{\text{BS}} \times N_{\text{MS}}$ precoder for the u th user.
 - \mathbf{D}_U satisfying the total TX power constraint $\|\mathbf{D}_u\|_F = N_{\text{MS}}$
- Considering the **narrowband block-fading channel**, the RX signal vector \mathbf{r}_u of the u th MS:

$$\mathbf{r}_u = \mathbf{H}_u \sum_{n=1}^U \mathbf{D}_n \mathbf{s}_n + \mathbf{n}_u$$

where

$\mathbf{s}_n : N_{\text{MS}} \times 1$ original signal vector with normalized power

$\mathbf{H}_u : N_{\text{MS}} \times N_{\text{BS}}$ mmWave massive MIMO matrix between BS and the u th MS

$\mathbf{n}_u : \text{iid AWGN noise } CN(0, \sigma_n^2)$

Multuser Digital Precoding

- The terms $\mathbf{H}_u \mathbf{D}_n \mathbf{s}_n$ for $n \neq u$ are interferences to the u th MS.
- Design the **BD precoding** \mathbf{D}_n to satisfy $\mathbf{H}_u \mathbf{D}_n = \mathbf{0}$. (Nulling)
- \mathbf{D}_n is designed to lie in the null space of $\bar{\mathbf{H}}_u$,

$$\begin{aligned} \bar{\mathbf{H}}_u &= \left[\mathbf{H}_1^H \cdots \mathbf{H}_{u-1}^H \quad \mathbf{H}_{u+1}^H \cdots \mathbf{H}_U^H \right]^H \\ &= \bar{\mathbf{U}}_u \bar{\mathbf{\Lambda}}_u \bar{\mathbf{V}}_u^H = \bar{\mathbf{U}}_u \bar{\mathbf{\Lambda}}_u \left[\bar{\mathbf{V}}_u^{nonzero}, \bar{\mathbf{V}}_u^{zero} \right]^H \end{aligned}$$

where

$\bar{\mathbf{V}}_u^{nonzero}$: The right singular vectors corresponding to nonzero singular values of $\bar{\mathbf{H}}_u$

$\bar{\mathbf{V}}_u^{zero}$: The right singular vectors corresponding to zero singular values of $\bar{\mathbf{H}}_u$

- The digital precoder \mathbf{D}_u of the u th MS is (first N_{MS} columns of $\bar{\mathbf{V}}_u^{zero}$)

$$\mathbf{D}_u = \bar{\mathbf{V}}_u^{zero}(:, 1:N_{MS})$$

Multiuser Digital Precoding

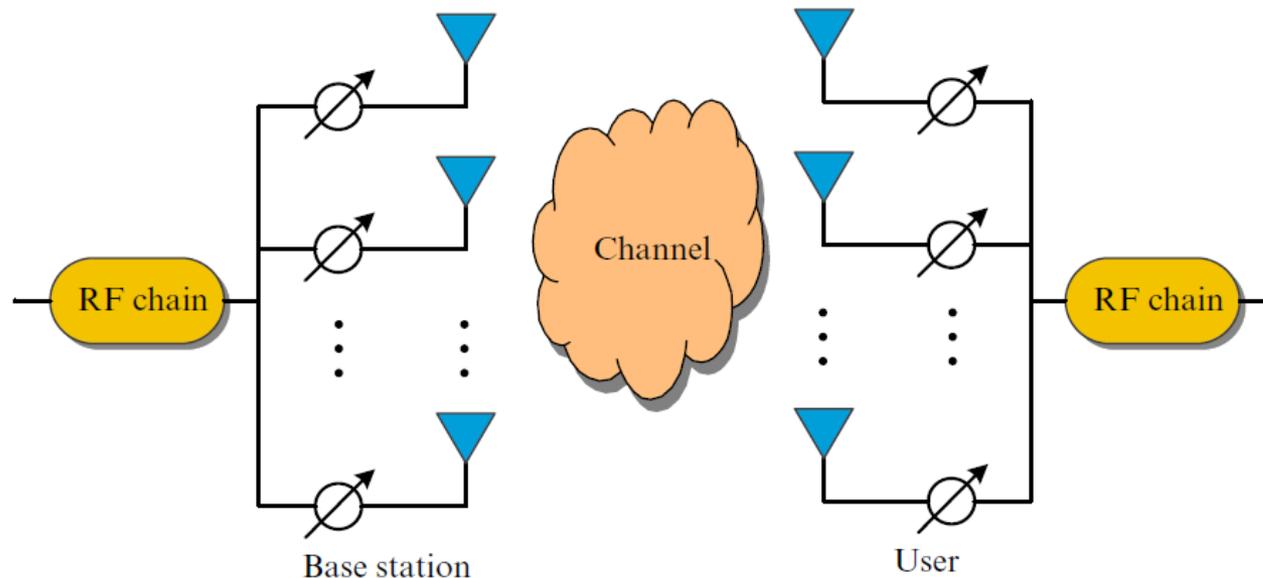
- Note: The optimal **DPC** and the near-optimal Tomlinson-Harashima (**TH**) precoding involve high computational complexity.

4. Analog Beamforming

- **Analog beamforming** is developed in point-to-point mmWave systems with large antenna arrays.
 - Only **one** RF chain
 - TX a **single** data stream
 - Control the **phases** of original signals to achieve the maximum antenna array gain and effective SNR.
- The widely used analog beamforming scheme:
 - **Beam steering**
 - To obtain the best analog beamforming vectors:
 - **Beam training schemes**

4.1 Beam Steering

- Architecture of analog beamforming for single-user mmWave massive MIMO Systems.



Architecture of analog beamforming for single-user mmWave massive MIMO systems.

- N_t TX antennas, N_r RX antennas, only one RF chain to TX one data stream.

Beam Steering

- Define $\mathbf{f}_{N_t \times 1}$ as BS analog **beamforming** vector, $\mathbf{w}_{N_r \times 1}$ as user analog **combining** vector.
- Design \mathbf{f} and \mathbf{w} to **maximize the effective SNR**:

$$(\mathbf{w}^{\text{opt}}, \mathbf{f}^{\text{opt}}) = \arg \max |\mathbf{w}^H \mathbf{H} \mathbf{f}|^2$$

$$\text{s.t. } w_i = \sqrt{N_r^{-1}} e^{j\phi_i}, \quad \forall i,$$

$$f_l = \sqrt{N_t^{-1}} e^{j\phi_l}, \quad \forall l.$$

- For the **optimal solution (unconstraint)**:

$$\mathbf{H} = \mathbf{U} \mathbf{\Sigma} \mathbf{V}^H \quad (\text{SVD of } \mathbf{H})$$

$$\mathbf{w}^{\text{opt}} = \mathbf{U}(:, 1) \quad (\text{Max. eig. value, eig. vector})$$

$$\mathbf{f}^{\text{opt}} = \mathbf{V}(:, 1) \quad (\text{No amplitude constraint})$$

Beam Steering

- Design the practical solutions \mathbf{f} and \mathbf{w} **satisfying the amplitude constraint** to be close to the optimal unconstrained solutions \mathbf{f}^{opt} and \mathbf{w}^{opt} .
 - Each **right** singular **vector** of \mathbf{H} with $L=o(N_t)$ converges in chordal distance to an array response vector $\mathbf{a}_t(\phi_\ell^t, \theta_\ell^t)$.
 - Each **left** singular **vector** of \mathbf{H} with $L=o(N_r)$ converges in chordal distance to an array response vector $\mathbf{a}_r(\phi_\ell^r, \theta_\ell^r)$.
 - Singular values converge to $N_t N_r |\alpha_\ell|^2 / L$.
- In other words, we can select

where
$$\mathbf{f} = \mathbf{a}_t(\phi_{k^*}^t, \theta_{k^*}^t) \text{ and } \mathbf{w} = \mathbf{a}_r(\phi_{k^*}^r, \theta_{k^*}^r)$$

$k^* = \arg \max_\ell |\alpha_\ell|^2$, to steer the beam in the strongest direction.

(For large N_t and N_r , it can achieve near-optimal performance)

4.2 Beam Training

- For the above beam steering, we need to **know the perfect CSI**
 - **Impractical** in the realistic systems due to only one RF chain
 - It observes a noisy version of the effective channel of smaller size.
- For the no full CSI, using the subspace sampling for beam training to find the \mathbf{f} and \mathbf{w} .
 - BS and User collaborate to **search the best beamformer** and combiner **pair** from the predefined codebooks during the beam training.

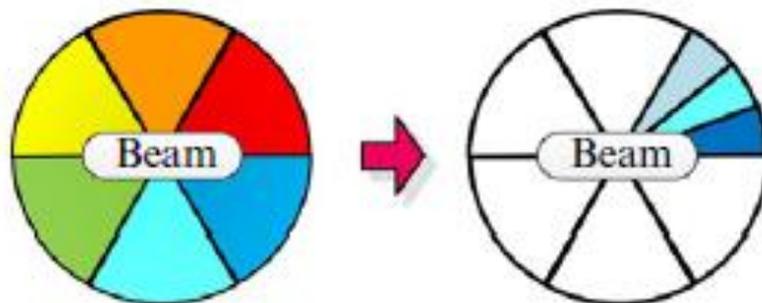
Beam Training

- The different **codebook sizes** are designed based on the beam steering scheme:

$$\mathbf{f} \in \mathcal{F} = \left\{ \mathbf{a}_t(\bar{\phi}_1^t, \bar{\theta}_1^t), \mathbf{a}_t(\bar{\phi}_2^t, \bar{\theta}_2^t), \dots, \mathbf{a}_t(\bar{\phi}_{|\mathcal{F}|}^t, \bar{\theta}_{|\mathcal{F}|}^t) \right\}$$

$$\mathbf{w} \in \mathcal{W} = \left\{ \mathbf{a}_r(\bar{\phi}_1^r, \bar{\theta}_1^r), \mathbf{a}_r(\bar{\phi}_2^r, \bar{\theta}_2^r), \dots, \mathbf{a}_r(\bar{\phi}_{|\mathcal{W}|}^r, \bar{\theta}_{|\mathcal{W}|}^r) \right\}$$

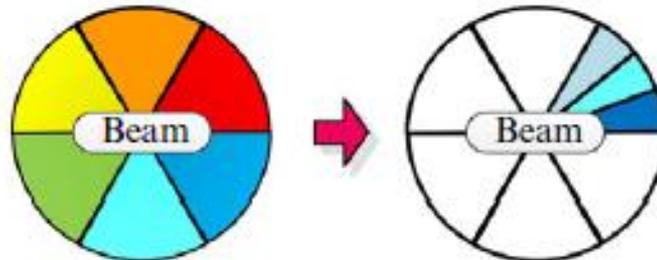
- It can uniformly cover the whole range of AoDs/AoAs.



(A) Multilevel codebook

Beam Training

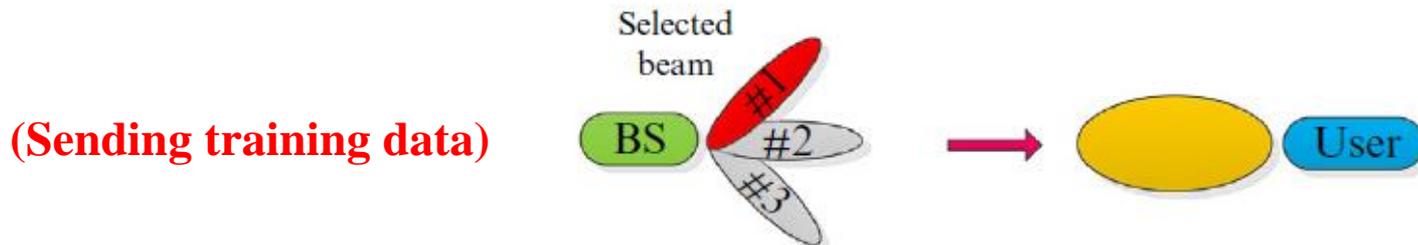
- For the **optimal** beam training scheme:
 - It is to **exhaustively search** all possible $|F| |W|$ pairs of beamforming and combining based on the maximized SNR criterion.
 - It cannot be affordable due to very large $|F|$ and $|W|$.
- For the **hierarchical beam training** scheme: (reducing the overhead of the exhaustive search)
 - (1) Construct a series of codebooks $F_1 F_2 \cdots F_K$ ($W_1 W_2 \cdots W_K$) with the **increasing resolution**.



(A) Multilevel codebook

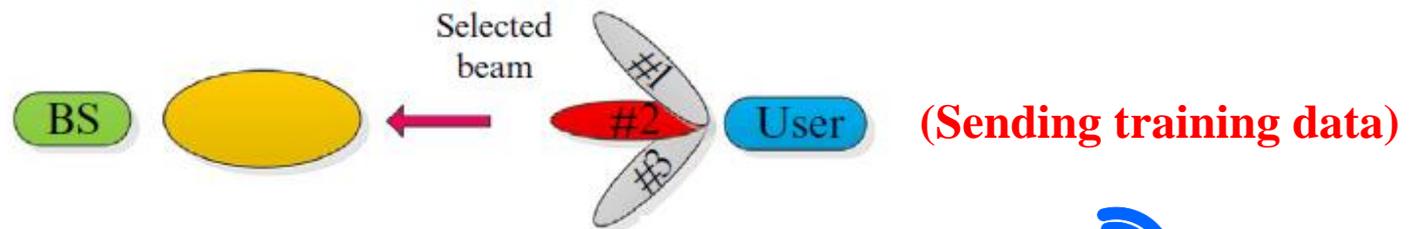
Beam Training

- (2) At the first level (lowest resolution codebook F_1), beam sweep at BS side (MS only RX and find the index of selected beamforming)



(B) Beam sweep at BS side

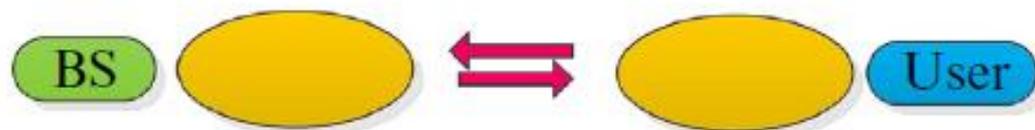
- (3) Swap their roles and beam sweep at user side to find the best beamforming index.



(C) beam sweep at user side

Beam Training

- (4) **Feedback the index** of the selected beamforming vector to each other.



(D) feedback phase

- (5) **Repeat** the above procedure with a **higher resolution codebook** within the chosen beam **until the last level** (highest resolution codebook, F_K)

5. Hybrid Precoding

- Hybrid (analog and digital) Precoding used in mmWave massive MIMO systems
 - Step1: a small-size digital precoder used to cancel interferences.
 - Step2: a large-size analog beamformer used to increase antenna array gain.
- Two architectures of hybrid beamforming:
 - (1) Fully connected architecture: each RF chain connected to all BS antennas via PSs.
 - (2) Subconnected architecture: each RF chain connected to only a subset of BS antennas.

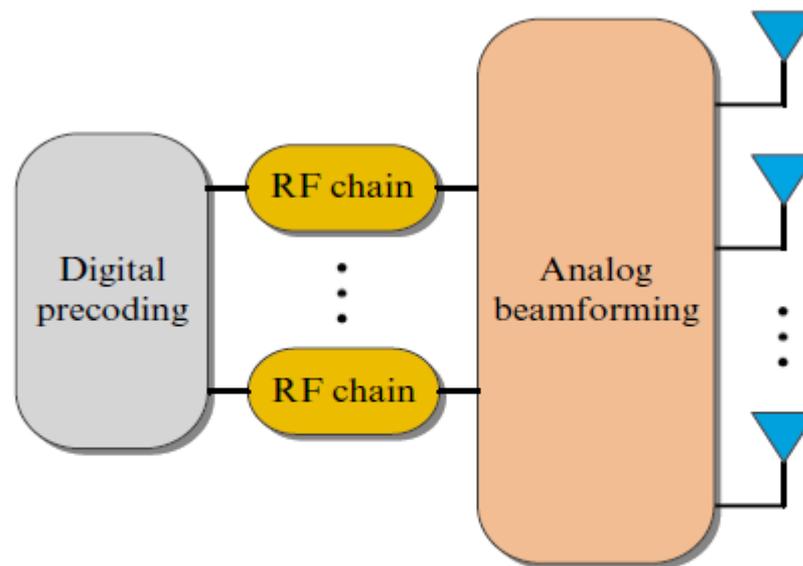
Hybrid Precoding

- Two systems of hybrid precoding:
 - (1) Single-user hybrid precoding
 - (2) Multiuser hybrid precoding
- Two hybrid beamforming schemes of single-user system:
 - (1) Spatially sparse hybrid precoding (fully connected architecture)
 - (2) Successive interference cancellation (SIC)-based hybrid precoding (subconnected architecture)
- Two-stage hybrid precoding scheme is used for multiuser system.

5.1 Single-User Hybrid Precoding

<System Model>

- Hardware architecture of hybrid precoding for single-user mmWave massive MIMO systems



Hardware architecture of hybrid precoding for single-user mmWave massive MIMO systems.

Single-User Hybrid Precoding

- .. N_t TX antennas, TX N_s data streams, a user with N_r antennas
- .. N_t^{RF} RF chains to TX multistream, $N_s \leq N_t^{RF} \leq N_t$
- .. BS applies an $N_t^{RF} \times N_s$ digital precoder D using N_t^{RF} RF chains
and an $N_t \times N_t^{RF}$ analog beamformer A using analog phase shifters(PSs)

- The transmitted signal:

$$\mathbf{x} = \mathbf{A}\mathbf{D}\mathbf{s}$$

where $\mathbf{s} : N_s \times 1$ original signal vector with normalized power $E(\mathbf{s}\mathbf{s}^H) = \frac{1}{N_s} \mathbf{I}_{N_s}$

- Consider a simple narrowband system:
 - Coherence bandwidth is usually very large at mmWave , e.q. the order of 100MHz.

Single-User Hybrid Precoding

- The received signal $\mathbf{y}_{N_r \times 1}$:

$$\mathbf{y} = \sqrt{\rho} \mathbf{H} \mathbf{A} \mathbf{D} \mathbf{s} + \mathbf{n}.$$

where

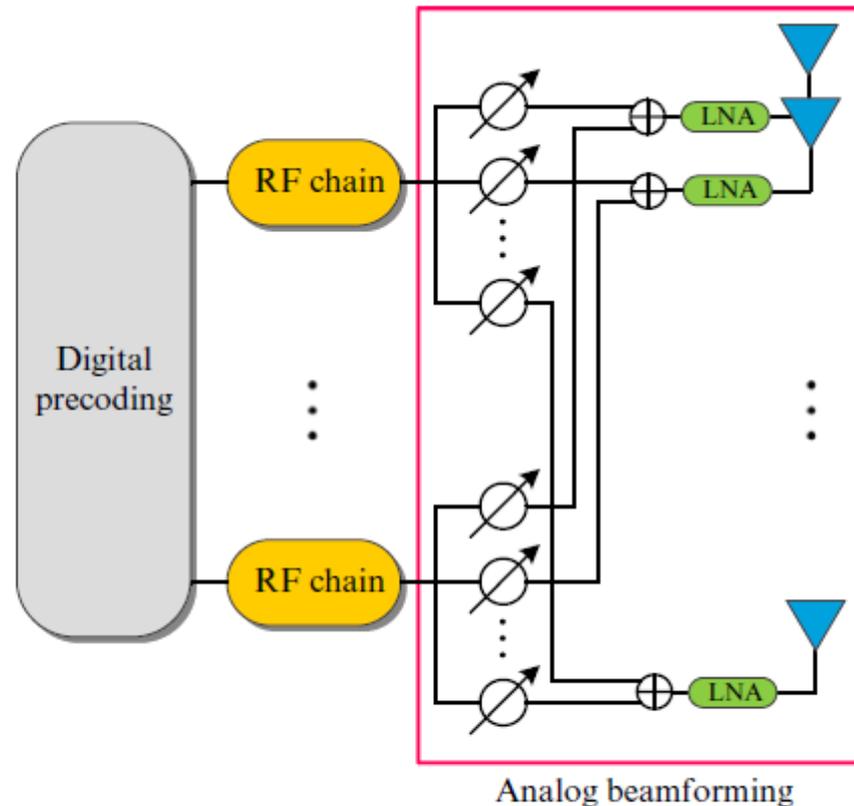
\mathbf{H} : *known at BS and user*

CSI: obtained at RX via training and shared with TX via limited feedback

<Spatially Sparse Hybrid Precoding (full connected)>

- Fully connected architecture of hybrid precoding for single-user mmWave massive MIMO system.

Single-User Hybrid Precoding



Fully connected architecture of hybrid precoding for single-user mmWave massive MIMO system.

Single-User Hybrid Precoding

- All elements of analog beamformer \mathbf{A} have the same amplitude $\frac{1}{\sqrt{N_t}}$ but the different phase shifters.
 - Total TX power constraint via the normalizing \mathbf{D} to satisfy $\|\mathbf{AD}\|_F^2 = N_s$
- Design (\mathbf{A}, \mathbf{D}) to maximize the sum rate $R(\mathbf{A}, \mathbf{D})$ achieved by Gaussian signaling over mmWave channel:

$$R(\mathbf{A}, \mathbf{D}) = \log_2 \left(\left| \mathbf{I} + \frac{\rho}{N_s \sigma_n^2} \mathbf{H} \mathbf{A} \mathbf{D} \mathbf{D}^H \mathbf{A}^H \mathbf{H}^H \right| \right).$$

- The corresponding sum-rate optimization problem:

$$\begin{aligned} (\mathbf{A}^{\text{opt}}, \mathbf{D}^{\text{opt}}) &= \arg \max_{\mathbf{A}, \mathbf{D}} R(\mathbf{A}, \mathbf{D}), \\ \text{s.t. } \mathbf{A} &\in \mathcal{F}, \\ \|\mathbf{AD}\|_F^2 &= N_s, \end{aligned}$$

Single-User Hybrid Precoding

where F : the set containing all feasible analog beamformers

the set of $N_t \times N_t^{RF}$ matrices with constant-magnitude entries

- It is known that there are no general solution due to the **nonconvex** amplitude constraint $\mathbf{A} \in F$
 - To find practical solutions using an approximation scheme, e.g., transforming the sum rate into “distance” between $\mathbf{A}\mathbf{D}$ and optimal unconstrained precoder \mathbf{P}_{opt} .
- First , using the SVD of $\mathbf{H} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^H$, the sum rate:

$$R(\mathbf{A}, \mathbf{D}) = \log_2 \left(\left| \mathbf{I} + \frac{\rho}{N_s \sigma_n^2} \mathbf{\Sigma}^2 \mathbf{V}^H \mathbf{A} \mathbf{D} \mathbf{D}^H \mathbf{A}^H \mathbf{V} \right| \right)$$

where $\mathbf{\Sigma} = \begin{bmatrix} \mathbf{\Sigma}_1 & \mathbf{0} \\ \mathbf{0} & \mathbf{\Sigma}_2 \end{bmatrix}$, $\mathbf{V} = [\mathbf{V}_1 \ \mathbf{V}_2]$, $\mathbf{\Sigma}_1 : N_s \times N_s$
 $\mathbf{V}_1 : N_t \times N_s$

Single-User Hybrid Precoding

\Rightarrow the optimal unconstrained precoder $\mathbf{P}_{opt} = \mathbf{V}_1$,

it cannot be expressed by AD with $\mathbf{A} \in \mathcal{F}$

- Second, the practical hybrid precoder AD is designed to close the optimal unconstrained precoder \mathbf{V}_1 . (Get near-optimal performance)

$\Rightarrow \max \text{tr}(\mathbf{V}_1^H \mathbf{A} \mathbf{D})$

\Rightarrow equivalent to $\min \|\mathbf{P}_{opt} - \mathbf{A} \mathbf{D}\|_F$

- Third, the near-optimal sum-rate approximated by

$$(\mathbf{A}^{opt}, \mathbf{D}^{opt}) = \arg \min_{\mathbf{A} \mathbf{D}} \|\mathbf{P}_{opt} - \mathbf{A} \mathbf{D}\|_F$$

$$\text{s.t. } \mathbf{A}(:, i) \in \{\mathbf{a}_t(\phi_\ell^t, \theta_\ell^t), \forall \ell\}$$

$$\|\mathbf{A} \mathbf{D}\|_F^2 = N_s$$

Single-User Hybrid Precoding

- Fourth, finding the best low-dimensional representation of \mathbf{P}_{opt} using the basis vector $\mathbf{a}_t(\phi_\ell^t, \theta_\ell^t)$.
 - Selecting the “best” N_t^{RF} array response vectors and finding their optimal baseband combination, the optimal objective function:

$$\tilde{\mathbf{D}}^{\text{opt}} = \arg \min_{\tilde{\mathbf{D}}} \left\| \mathbf{P}_{\text{opt}} - \mathbf{A}_t \tilde{\mathbf{D}} \right\|_F,$$

$$\text{s.t. } \left\| \text{diag}(\tilde{\mathbf{D}} \tilde{\mathbf{D}}^H) \right\|_0 = N_t^{\text{RF}},$$

$$\left\| \mathbf{A}_t \tilde{\mathbf{D}} \right\|_F^2 = N_s,$$

where

$$\mathbf{A}_t = \left[\mathbf{a}_t(\phi_1^t, \theta_1^t), \dots, \mathbf{a}_t(\phi_L^t, \theta_L^t) \right] \Rightarrow L \leq N_t^{\text{RF}}$$

$$\tilde{\mathbf{D}}: L \times N_s \text{ matrix, } \left\| \text{diag}(\tilde{\mathbf{D}} \tilde{\mathbf{D}}^H) \right\|_0 = N_t^{\text{RF}}, L \leq N_t^{\text{RF}} \text{ (select } N_t^{\text{RF}} \text{ rows of } \tilde{\mathbf{D}})$$

Single-User Hybrid Precoding

- It is equivalent to the typical problem of sparse signal recovery. The above problem can be solved by the well-known concept of **orthogonal matching pursuit** (OMP).

ALGORITHM 1 SPATIALLY SPARSE PRECODING

Input: \mathbf{P}_{opt}

- 1: $\mathbf{A} = \text{Empty Matrix}$
- 2: $\mathbf{P}_{\text{res}} = \mathbf{P}_{\text{opt}}$
- 3: **for** $i \leq N_t^{\text{RF}}$ **do**
- 4: $\mathbf{\Psi} = \mathbf{A}_t^H \mathbf{P}_{\text{res}}$
- 5: $k = \arg \max_{l=1, \dots, L} (\mathbf{\Psi} \mathbf{\Psi}^H)_{l,l}$
- 6: $\mathbf{A} = \left[\mathbf{A} \mid \mathbf{A}_t^{(k)} \right]$
- 7: $\mathbf{D} = (\mathbf{A}^H \mathbf{A})^{-1} \mathbf{A}^H \mathbf{P}_{\text{opt}}$
- 8: $\mathbf{P}_{\text{res}} = \frac{\mathbf{P}_{\text{opt}} - \mathbf{A} \mathbf{D}}{\|\mathbf{P}_{\text{opt}} - \mathbf{A} \mathbf{D}\|_F}$
- 9: **end for**
- 10: $\mathbf{D} = \sqrt{N_s} \frac{\mathbf{D}}{\|\mathbf{A} \mathbf{D}\|_F}$
- 11: **return** \mathbf{A}, \mathbf{D}

Single-User Hybrid Precoding

- In Algorithm1:
 - (1) Step1&2: Initialization
 - (2) Step 5: Finding the vector $\mathbf{a}_t(\phi_\ell^t, \theta_\ell^t)$, which the optimal precoder has the maximum projection.
 - (3) Step 6: Appending the selected $\mathbf{a}_t(\phi_\ell^t, \theta_\ell^t)$ to the analog beamformer \mathbf{A} .
 - (4) Step 7: LS solution \mathbf{D} is calculated by the dominant \mathbf{A}
 - (5) Step 8: Removing the dominant \mathbf{A} contribution, the residual precoding matrix \mathbf{P}_{res} to find the next \mathbf{a}_t and get largest projection.
 - (6) Step 9: Until all N_t^{RF} precoding vectors have been selected, the \mathbf{A} and \mathbf{D} are determined to $\min \|\mathbf{P}_{\text{opt}} - \mathbf{AD}\|_F$

Single-User Hybrid Precoding

(7) Step 10: Ensuring to satisfy the TX power constraint $\|\mathbf{AD}\|_F^2 = N_s$

Single-User Hybrid Precoding

<Performance Evaluation>

□ Simulation parameters:

■ Three methods comparison:

□ (1). Spatially sparse precoding

□ (2). Optimal unconstrained precoding ($\mathbf{P}_{opt} = \mathbf{V}_1$, fully digital precoding)

□ (3). Beam steering precoding (fully analog beamforming)

■ $L=3$ propagation paths with uniformly distributed AoAs/AoDs

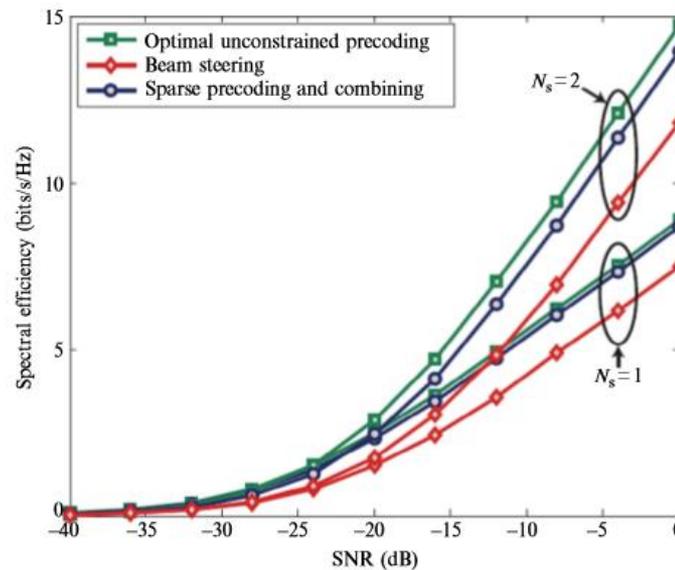
■ BS sector angle: AZ=60°, EL=20° wide

MS: smaller antenna arrays of omnidirectional elements (steer any direction)

■ Element spacing $d = \lambda/2$, $SNR = \rho/\sigma_n^2$

Single-User Hybrid Precoding

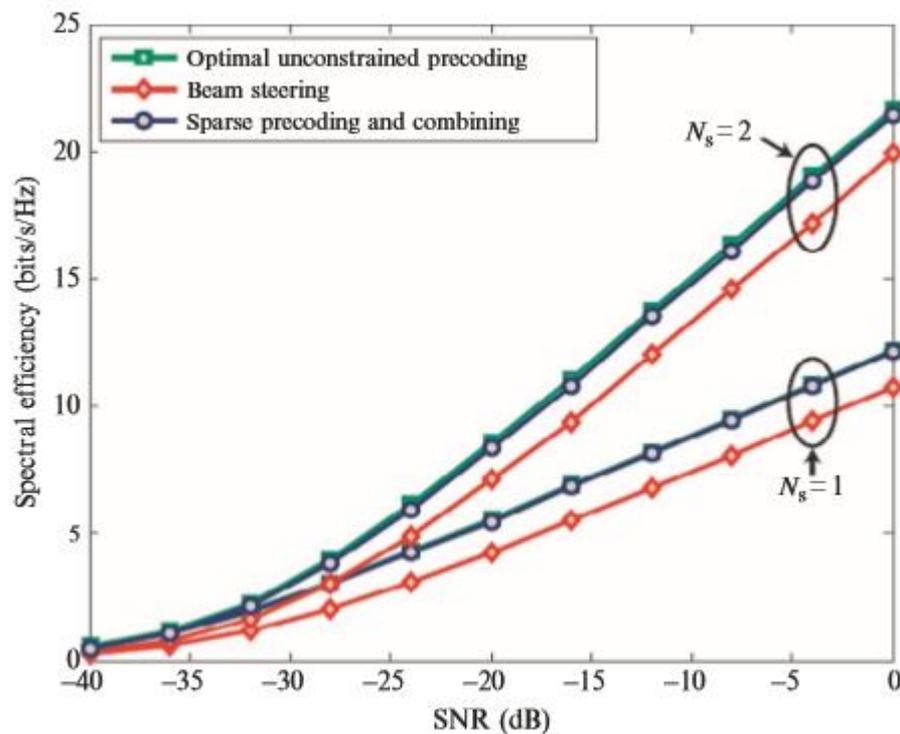
- 64x16 mmWave massive MIMO system with planar arrays at BS and MS
- BS uses $N_t^{RF} = 4$ RF chains to TX $N_s = 1$ or 2 streams
- Simulation results:
 - Sum-rate comparison in 64x16 mmWave massive MIMO system with $N_t^{RF} = 4$



Sum-rate comparison in a 64×16 mmWave massive MIMO system with $N_t^{RF} = 4$.

Single-User Hybrid Precoding

- Sum-rate comparison in 256x64 mmWave massive MIMO system with $N_t^{RF} = 6$



$$N_t^{RF} \uparrow$$

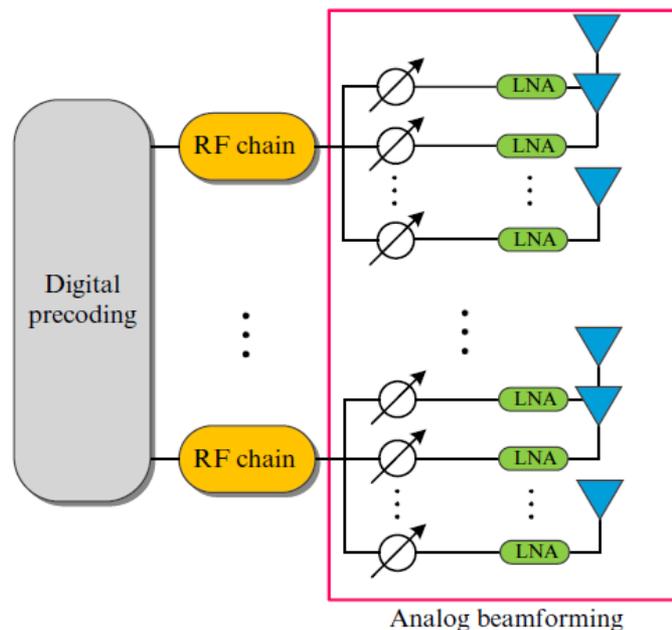
Sparse precoding close to optimal precoding

Sum-rate comparison in a 256 × 64 mmWave massive MIMO system with $N_t^{RF} = 6$.

Single-User Hybrid Precoding

SIC-Based Hybrid Precoding (Subconnected)

- For the subconnected architecture, each RF chain is connected to only a subset of BS antennas.



Subconnected architecture of hybrid precoding for single-user mmWave massive MIMO system.

- Reducing the number of required PSs from $N_t \times N_t^{RF}$ to N_t
- low computation complexity

SIC-Based Hybrid Precoding (Subconnected)

- Consider the single-user mmWave massive MIMO system:
 - BS with N_t antennas and N_t^{RF} RF chains
 - Each RF chain connected to one subantenna array with M antennas, i.e., $N_t = N_t^{RF} M$
 - BS TX $N_s = N_t^{RF}$ streams (fully spatial multiplexing gain)
 - MS user with N_r receiver antennas
 - The digital precoder $\mathbf{D}_{N_t^{RF} \times N_s = N_s \times N_s}$ being specialized to a diagonal matrix $\mathbf{D} = \text{diag}[d_1 \ d_2 \ \cdots \ d_{N_s}]$ (\mathbf{D} : power allocation)

SIC-Based Hybrid Precoding (Subconnected)

- The analog beamformer $\mathbf{A}_{N_t \times N_t^{RF} = N_t \times N_s}$ is a special block diagonal structure

$$\mathbf{A} = \begin{bmatrix} \bar{\mathbf{a}}_1 & \mathbf{0} & \dots & \mathbf{0} \\ \mathbf{0} & \bar{\mathbf{a}}_2 & & \mathbf{0} \\ \vdots & & \ddots & \vdots \\ \mathbf{0} & \mathbf{0} & \dots & \bar{\mathbf{a}}_{N_s} \end{bmatrix}_{N_t \times N_s}, \quad \bar{\mathbf{a}}_n \in \mathbb{C}^{M \times 1}$$

where the elements $\bar{\mathbf{a}}_n$ have the same amplitude $\frac{1}{\sqrt{M}}$ but different phases.

SIC-Based Hybrid Precoding (Subconnected)

<Basic Idea>

- Maximizing the total achievable rate $R(\mathbf{P})$ to design the hybrid precoder $\mathbf{P}=\mathbf{A}\mathbf{D}$

$$R(\mathbf{P}) = \log_2 \left(\left| I + \frac{\rho}{N_s \sigma_n^2} \mathbf{H} \mathbf{P} \mathbf{P}^H \mathbf{H}^H \right| \right)$$

where

$$\mathbf{P} = \mathbf{A}\mathbf{D} = \text{diag} \{ \bar{\mathbf{a}}_1, \dots, \bar{\mathbf{a}}_{N_s} \} \cdot \text{diag} \{ d_1, \dots, d_{N_s} \}$$

SIC-Based Hybrid Precoding (Subconnected)

□ Satisfying the three constraints:

(1) $\mathbf{P} = \text{diag} \{ \bar{\mathbf{p}}_1, \dots, \bar{\mathbf{p}}_{N_s} \}$, $\bar{\mathbf{p}}_n = d_n \bar{\mathbf{a}}_n$ is the $M \times 1$ vector of \mathbf{p}_n of \mathbf{P}

$$\mathbf{p}_n = [\mathbf{0}_{1 \times M(n-1)}, \bar{\mathbf{p}}_n^T, \mathbf{0}_{1 \times M(N_s-n)}]^T$$

(2) The amplitude of the elements of \mathbf{A} is fixed to $\frac{1}{\sqrt{M}}$

The elements of each column of \mathbf{P} have the same amplitude due to the diagonal \mathbf{D} .

(3) $\|\mathbf{P}\|_F \leq N_s$ meeting the total TX power constraint

→ (1) & (2) nonconvex constraints

SIC-Based Hybrid Precoding (Subconnected)

- Based on the special block diagonal hybrid precoding \mathbf{P} , the precoding on different subarrays is independent.
 - Decompose the total sum rate into a series of subrate optimization problems, each subrate only considers one subarray.
- Diving \mathbf{P} as $\mathbf{P} = \begin{bmatrix} \mathbf{P}_{N_s-1} & \mathbf{p}_{N_s} \end{bmatrix}$, \mathbf{p}_{N_s} is the N_s th columns of \mathbf{P} , \mathbf{P}_{N_s-1} is the first $(N_s - 1)$ columns of \mathbf{P} . ($N_s M \times (N_s - 1)$)

The total achievable rate $R(\mathbf{P})$:

$$R(\mathbf{P}) \stackrel{(a)}{=} \log_2(|\mathbf{T}_{N_s-1}|) + \log_2 \left(\left| \mathbf{I} + \frac{\rho}{N_s \sigma_n^2} \mathbf{T}_{N_s-1}^{-1} \mathbf{H} \mathbf{p}_{N_s} \mathbf{p}_{N_s}^H \mathbf{H}^H \right| \right)$$

$$\stackrel{(b)}{=} \log_2(|\mathbf{T}_{N_s-1}|) + \log_2 \left(\left| \mathbf{I} + \frac{\rho}{N_s \sigma_n^2} \mathbf{p}_{N_s}^H \mathbf{H}^H \mathbf{T}_{N_s-1}^{-1} \mathbf{H} \mathbf{p}_{N_s} \right| \right)$$

SIC-Based Hybrid Precoding (Subconnected)

where

$$(a) : \mathbf{T}_{N_s-1} = \mathbf{I} + \frac{\rho}{N_s \sigma_n^2} \mathbf{H} \mathbf{P}_{N_s-1} \mathbf{P}_{N_s-1}^H \mathbf{H}^H$$

$$(b) : |\mathbf{I} + \mathbf{X}\mathbf{Y}| = |\mathbf{I} + \mathbf{Y}\mathbf{X}|, \mathbf{X} = \mathbf{T}_{N_s-1}^{-1} \mathbf{H} \mathbf{p}_{N_s}, \mathbf{Y} = \mathbf{p}_{N_s}^H \mathbf{H}^H$$

The right side of (b) is the achievable subrate of the N_s th subarray.

→ Next, further decomposing $\log_2 \left(|\mathbf{T}_{N_s-1}| \right)$:

$$\log_2 \left(|\mathbf{T}_{N_s-2}| \right) + \log_2 \left(\mathbf{I} + \frac{\rho}{N_s \sigma_n^2} \mathbf{p}_{N_s-1}^H \mathbf{H}^H \mathbf{T}_{N_s-2}^{-1} \mathbf{H} \mathbf{p}_{N_s-1} \right)$$

SIC-Based Hybrid Precoding (Subconnected)

- After N decompositions, the total achievable rate R :

$$R = \sum_{n=1}^{N_s} \log_2 \left(I + \frac{\rho}{N_s \sigma_n^2} \mathbf{p}_n^H \mathbf{H}^H \mathbf{T}_{n-1}^{-1} \mathbf{H} \mathbf{p}_n \right)$$

$$\mathbf{T}_n = I + \frac{\rho}{N_s \sigma_n^2} \mathbf{H} \mathbf{P}_n \mathbf{P}_n^H \mathbf{H}^H, \quad \mathbf{T}_0 = \mathbf{I}_{N_s}$$

→ Optimal the subrate of 1st subantenna and update the \mathbf{T}_1 matrix, until the last subarray, i.e., SIC-based hybrid precoding

SIC-Based Hybrid Precoding (Subconnected)

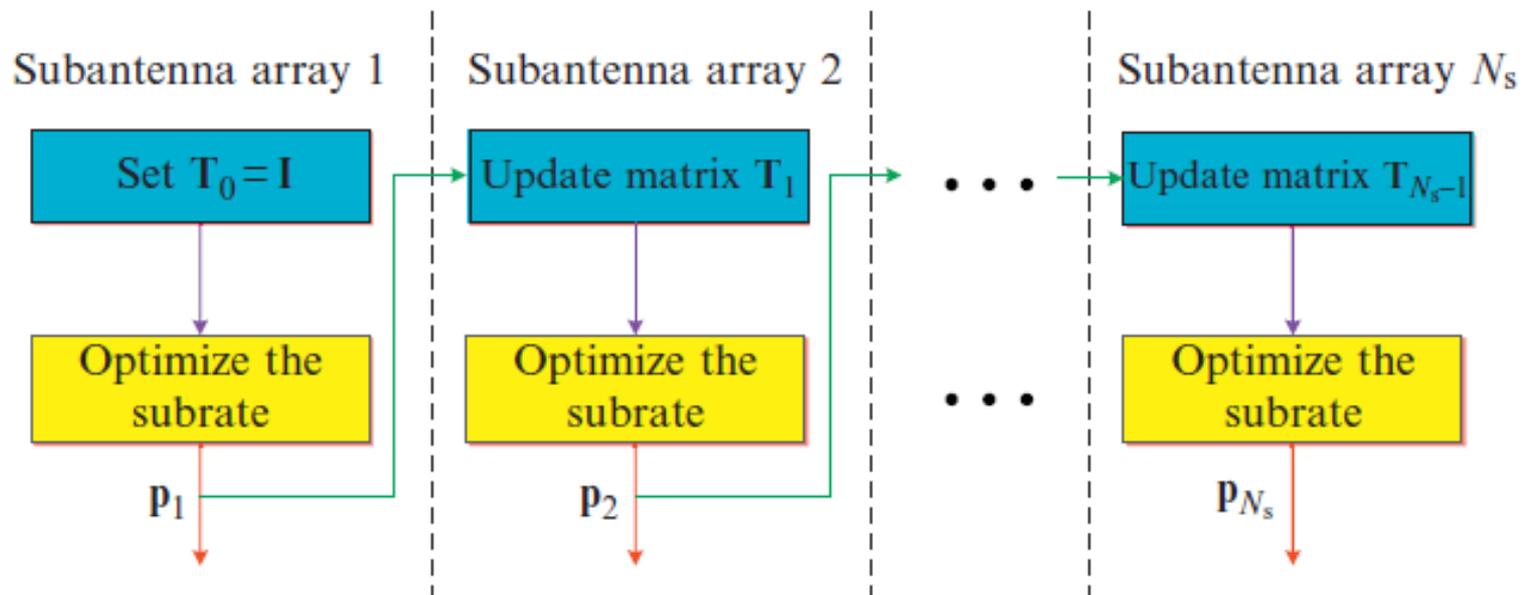


Diagram of the SIC-based hybrid precoding.

SIC-Based Hybrid Precoding (Subconnected)

- Focusing on subrate optimization problem of the n th subarray:

$$\mathbf{p}_n^{\text{opt}} = \arg \max_{\mathbf{p}_n \in \mathcal{F}} \log_2 \left(1 + \frac{\rho}{N_s \sigma_n^2} \mathbf{p}_n^H \mathbf{G}_{n-1} \mathbf{p}_n \right),$$

where

$$\mathbf{G}_{n-1} = \mathbf{H}^H \mathbf{T}_{n-1}^{-1} \mathbf{H}$$

\mathcal{F} is the set of the feasible vectors satisfying three constraints.

- \mathbf{p}_n only has M nonzero elements $\bar{\mathbf{p}}_n$, the subrate optimization problem:

$$\bar{\mathbf{p}}_n^{\text{opt}} = \arg \max_{\bar{\mathbf{p}}_n \in \bar{\mathcal{F}}} \log_2 \left(1 + \frac{\rho}{N_s \sigma_n^2} \bar{\mathbf{p}}_n^H \bar{\mathbf{G}}_{n-1} \bar{\mathbf{p}}_n \right)$$

SIC-Based Hybrid Precoding (Subconnected)

□ where

$$\bar{\mathbf{G}}_{n-1} (\text{Submatrix of } \mathbf{G}_n) = \mathbf{R}\mathbf{G}_{n-1}\mathbf{R}^H = \mathbf{R}\mathbf{H}^H\mathbf{T}_{n-1}^{-1}\mathbf{H}\mathbf{R}^H$$

$$\mathbf{R} = \begin{bmatrix} \mathbf{0}_{M \times M(n-1)} & \mathbf{I}_M & \mathbf{0}_{M \times M(N_s-n)} \end{bmatrix}$$

$\bar{\mathbf{F}}$ is the $M \times 1$ feasible vectors satisfying constraints 2 & 3

→ $\bar{\mathbf{G}}_{n-1} = \mathbf{V}\mathbf{\Sigma}\mathbf{V}^H$, the optimal unconstrained precoding vector:

$$\bar{\mathbf{p}}_n^{opt} = \mathbf{v}_1 \text{ (the first column of } \mathbf{V}\text{)}$$

(Not obey the constraint 2, same amplitude)

SIC-Based Hybrid Precoding (Subconnected)

- Convert to the distance optimization problem : (equivalent)

$$\bar{\mathbf{p}}_n^{\text{opt}} = \arg \min_{\bar{\mathbf{p}}_n \in \bar{\mathcal{F}}} \|\mathbf{v}_1 - \bar{\mathbf{p}}_n\|_2^2$$

- The distance can be expressed by : (Using $\bar{\mathbf{p}}_n = d_n \bar{\mathbf{a}}_n$ constraints)

$$\begin{aligned} \|\mathbf{v}_1 - \bar{\mathbf{p}}_n\|_2^2 &= 1 + d_n^2 - 2d_n \operatorname{Re}(\mathbf{v}_1^H \bar{\mathbf{a}}_n) \\ &= (d_n - \operatorname{Re}(\mathbf{v}_1^H \bar{\mathbf{a}}_n))^2 + (1 - [\operatorname{Re}(\mathbf{v}_1^H \bar{\mathbf{a}}_n)]^2) \end{aligned}$$

\uparrow
 1st min.

\uparrow
 2nd min.

SIC-Based Hybrid Precoding (Subconnected)

where $\mathbf{v}_1^H \mathbf{v}_1 = 1$, $\bar{\mathbf{a}}_n^H \bar{\mathbf{a}}_n = 1$, each elements $\bar{\mathbf{a}}_n$ has the same amplitude $\frac{1}{\sqrt{M}}$

$$\rightarrow 1^{\text{st}} \text{ min.} : d_n^{\text{opt}} = \text{Re}(\mathbf{v}_1^H \bar{\mathbf{a}}_n)$$

$$2^{\text{nd}} \text{ min.} = \max \left| \text{Re}(\mathbf{v}_1^H \bar{\mathbf{a}}_n) \right| \quad (\because \mathbf{v}_1^H \mathbf{v}_1 = 1, \bar{\mathbf{a}}_n^H \bar{\mathbf{a}}_n = 1)$$

$$\therefore \bar{\mathbf{a}}_n^{\text{opt}} = \frac{1}{\sqrt{M}} e^{j\text{angle}(\mathbf{v}_1)} \quad (\text{elements of } \bar{\mathbf{a}}_n \text{ with same amplitude})$$

$$\therefore \text{get } d_n^{\text{opt}} :$$

$$d_n^{\text{opt}} = \text{Re}(\mathbf{v}_1^H \bar{\mathbf{a}}_n) = \frac{1}{\sqrt{M}} \cdot \text{Re}(\mathbf{v}_1^H e^{j\text{angle}(\mathbf{v}_1)}) = \frac{\|\mathbf{v}_1\|_1}{\sqrt{M}}$$

SIC-Based Hybrid Precoding (Subconnected)

- Get the optimal $\bar{\mathbf{p}}_n^{\text{opt}}$

$$\bar{\mathbf{p}}_n^{\text{opt}} = d_n^{\text{opt}} \bar{\mathbf{a}}_n^{\text{opt}} = \frac{1}{M} \|\mathbf{v}_1\|_1 e^{j\text{angle}(\mathbf{v}_1)}$$

where each element v_i of \mathbf{v}_1 has amplitude less than one,
i.e. , $\|\bar{\mathbf{p}}_n^{\text{opt}}\|_2^2 \leq 1$

$$\therefore \|\mathbf{P}^{\text{opt}}\|_F^2 = \|\text{diag}\{\bar{\mathbf{p}}_1^{\text{opt}}, \dots, \bar{\mathbf{p}}_N^{\text{opt}}\}\|_F^2 \leq N_s \quad (\text{satisfying constraint 3})$$

TX total PWR

- After acquiring $\bar{\mathbf{p}}_n^{\text{opt}}$, \mathbf{T}_n and $\bar{\mathbf{G}}_n$ can be updated to find next $\bar{\mathbf{p}}_{n+1}^{\text{opt}}$

SIC-Based Hybrid Precoding (Subconnected)

- Summary the following three steps:

Step 1: Execute the SVD of $\bar{\mathbf{G}}_{n-1}$ to obtain \mathbf{v}_1 .

Step 2: Let $\bar{\mathbf{p}}_n^{\text{opt}} = \frac{1}{M} \|\mathbf{v}_1\|_1 e^{j\text{angle}(\mathbf{v}_1)}$ be the optimal solution to the current n th subantenna array.

Step 3: Update matrices $\mathbf{T}_n = \mathbf{I} + \frac{\rho}{N_s \sigma_n^2} \mathbf{H} \mathbf{P}_n \mathbf{P}_n^H \mathbf{H}^H$ and $\bar{\mathbf{G}}_n = \mathbf{R} \mathbf{H}^H \mathbf{T}_n^{-1} \mathbf{H} \mathbf{R}^H$ for the next $(n+1)$ th subantenna array.

SIC-Based Hybrid Precoding (Subconnected)

<Performance Evaluation>

□ Simulation parameters:

.. $L=3$ (channel paths)

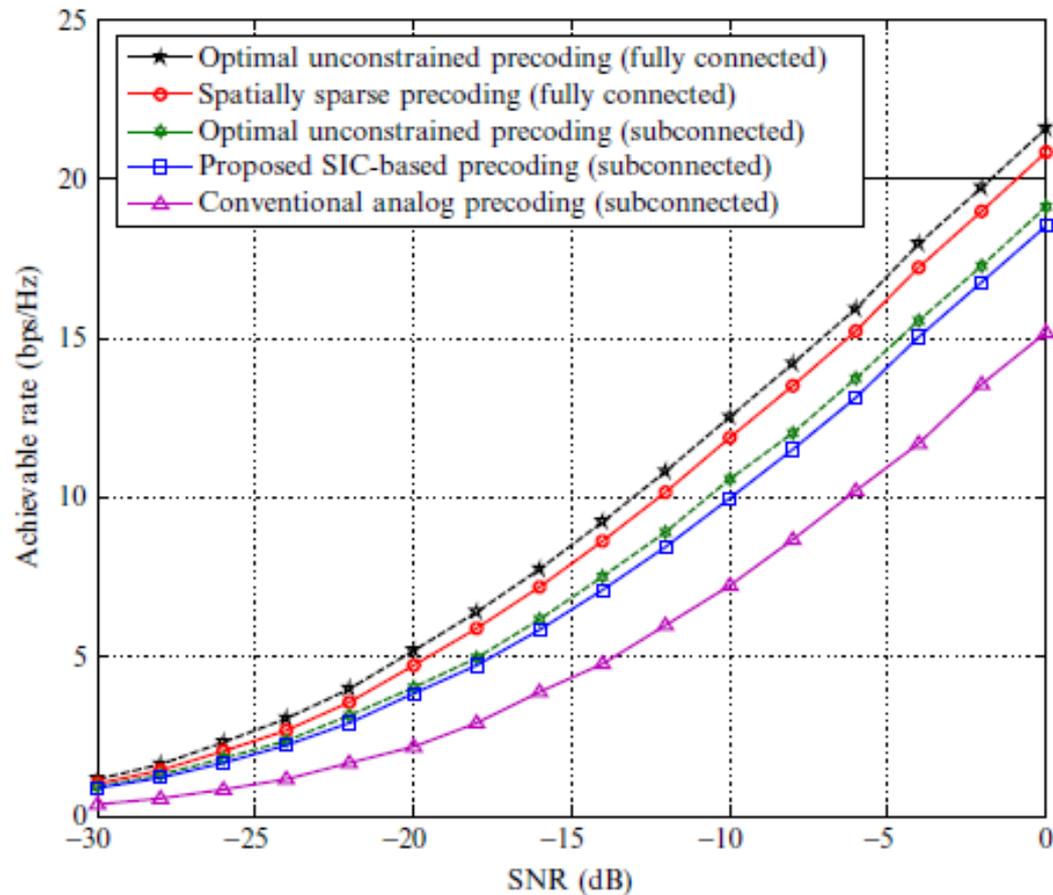
.. carrier frequency = 28GHz

.. $d = \lambda/2$

.. BS (AoDs = $[-\pi/6, \pi/6]$), MS (AoAs = $[-\pi, \pi]$)

.. $N_t \times N_r = 64 \times 16$ ($N_t^{\text{RF}} = N_s = 8$)

SIC-Based Hybrid Precoding (Subconnected)



Achievable rate comparison for an $N_t \times N_r = 64 \times 16$ ($N_t^{\text{RF}} = N_s = 8$) mmWave massive MIMO system.

SIC-Based Hybrid Precoding (Subconnected)

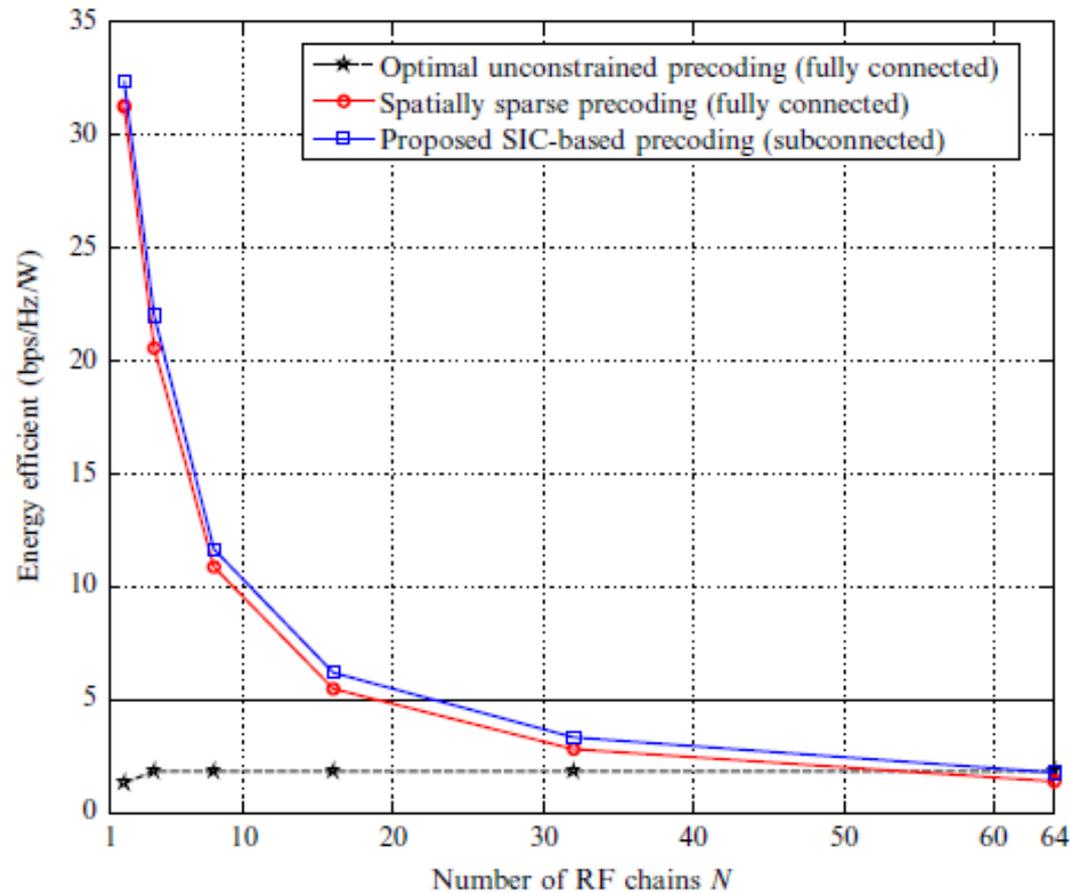
- Energy efficiency η :

$$\eta = \frac{R}{P_{\text{total}}} = \frac{R}{P_t + N_{\text{RF}}P_{\text{RF}} + N_{\text{PS}}P_{\text{PS}}} \text{ (bps/Hz/W)}$$

($P_{\text{RF}} = 250 \text{ mW}$, $P_{\text{PS}} = 1 \text{ mW}$, $P_t = 1 \text{ W (30 dBm)}$, $N_t \times N_r = 64 \times 64$)

($N_t^{\text{RF}} = 1, 2, 4, \dots, 64$)

SIC-Based Hybrid Precoding (Subconnected)

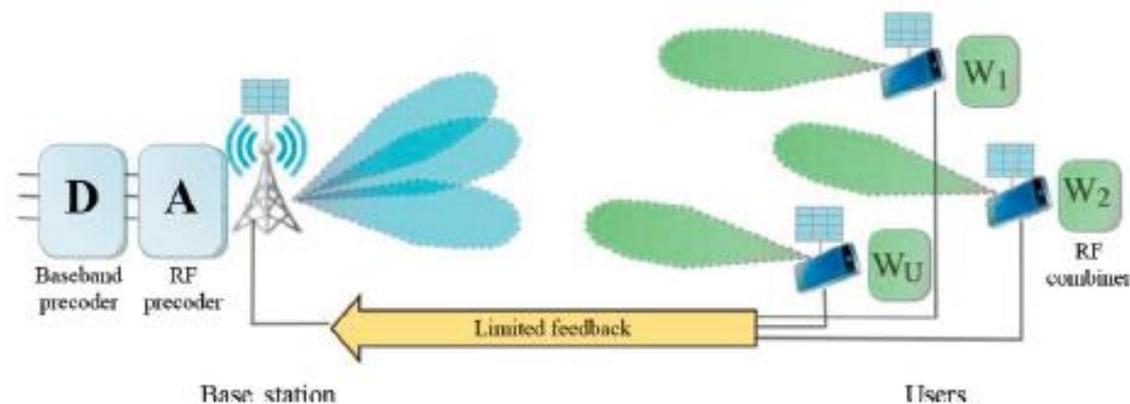


Energy efficiency comparison against the number of RF chains M_t^{RF} , where $M_t \times M_r = 64 \times 64$, SNR = 0 dB.

5.2 Multiuser Hybrid Precoding

<System Model>

- Hardware architecture of hybrid precoding for multiuser mmWave massive MIMO Systems



Hardware architecture of hybrid precoding for multiuser mmWave massive MIMO systems.

- BS : N_{BS} antennas, N_{RF} RF chains ($N_{RF} \leq N_{BS}$), communicating with U MSs
- Each MS: N_{MS} antennas, only one RF chain

Multuser Hybrid Precoding

- BS communicates with every MS via only one stream
- Total data streams , $N_S = U \leq N_{\text{RF}}$
- Fully connected architecture
- For the downlink , BS employs a $U \times U$ digital precoder $\mathbf{D} = [\mathbf{d}_1, \mathbf{d}_2, \dots, \mathbf{d}_U]$ and an $N_{\text{BS}} \times U$ analog precoder $\mathbf{A} = [\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_U]$, TX signal:

$$\mathbf{x} = \mathbf{A}\mathbf{D}\mathbf{s}$$

where $\underline{\mathbf{s}}_{U \times 1}$ original signal vector, $\mathbb{E}(\mathbf{s}\mathbf{s}^H) = (\rho/U)\mathbf{I}_U$,
 ρ : average TX power each user with equal power allocation

Multiuser Hybrid Precoding

all the elements of \mathbf{A} with the same constant amplitude N_{BS}^{-1}
the total TX power constraint with normalizing \mathbf{D} to satisfy
 $\|\mathbf{A}\mathbf{D}\|_F^2 = U$.

Multuser Hybrid Precoding

- Consider the narrowband block-fading channel model. The RX signal vector \mathbf{r}_u of the u th MS:

$$\mathbf{r}_u = \mathbf{H}_u \sum_{n=1}^U \mathbf{A} \mathbf{d}_n s_n + \mathbf{n}_u$$

where $\mathbf{H}_u : N_{\text{MS}} \times N_{\text{BS}}$ channel matrix between BS and the u th MS

$$\mathbf{n}_u : \mathcal{CN}(0, \sigma_n^2)$$

- Analog combiner \mathbf{w}_u of the u th MS is used to combine the RX signal

\mathbf{r}_u :

$$y_u = \mathbf{w}_u^H \mathbf{r}_u = \mathbf{w}_u^H \mathbf{H}_u \sum_{n=1}^U \mathbf{A} \mathbf{d}_n s_n + \mathbf{w}_u^H \mathbf{n}_u$$

Multiuser Hybrid Precoding

where \mathbf{w}_u is the same constraint as the analog precoder \mathbf{A} , i.e., all the elements of \mathbf{w}_u having the same amplitude N_{MS}^{-1} but different phases. (MS only analog BF, lower cost)

Multiuser Hybrid Precoding

<Two-Stage Hybrid Precoding>

- Goal : Designing BS analog precoder \mathbf{A} and digital precoder \mathbf{D} , MS analog combiner $\{\mathbf{w}_u\}_{u=1}^U$ to max sum rate:

$$R = \sum_{u=1}^U R_u$$

$$R_u = \log_2 \left(1 + \frac{\frac{P}{U} |\mathbf{w}_u^H \mathbf{H}_u \mathbf{A} \mathbf{d}_u|^2}{\frac{P}{U} \sum_{n \neq u} |\mathbf{w}_u^H \mathbf{H}_u \mathbf{A} \mathbf{d}_n|^2 + \sigma_n^2} \right)$$

Multiuser Hybrid Precoding

- Precoding design problem to find $\mathbf{A}^{\text{opt}}, \mathbf{D}^{\text{opt}}, \{\mathbf{w}_u^{\text{opt}}\}_{u=1}^U$ that solve

$$\left\{ \mathbf{A}^{\text{opt}}, \mathbf{D}^{\text{opt}}, \{\mathbf{w}_u^{\text{opt}}\}_{u=1}^U \right\} = \arg \max \sum_{u=1}^U \log_2 \left(1 + \frac{\frac{P}{U} |\mathbf{w}_u^H \mathbf{H}_u \mathbf{A} \mathbf{d}_u|^2}{\frac{P}{U} \sum_{n \neq u} |\mathbf{w}_u^H \mathbf{H}_u \mathbf{A} \mathbf{d}_n|^2 + \sigma_n^2} \right)$$

s.t. $\mathbf{a}_u \in \mathcal{F}, u = 1, 2, \dots, U,$
 $\mathbf{w}_u \in \mathcal{W}, u = 1, 2, \dots, U,$
 $\|\mathbf{A} \mathbf{D}\|_F^2 = U.$

- Step 1: Search over the entire $\mathcal{F}^U \times \mathcal{W}^U$ of all possible $\{\mathbf{a}_u\}_{u=1}^U$ and $\{\mathbf{w}_u\}_{u=1}^U$ to max each user power.
- Step 2: Using iterative method to find the digital precoder \mathbf{D} and suppress interferences.

Multuser Hybrid Precoding

ALGORITHM 2 TWO-STAGE HYBRID PRECODING

Inputs: \mathcal{F} BS analog precoding codebook

\mathcal{W} MS analog combining codebook

First stage: Single-user analog precoding/combining design

For each MS u , $u = 1, 2, \dots, U$

The BS and MS u select $\mathbf{v}_u^{\text{opt}}$ and $\mathbf{g}_u^{\text{opt}}$ respectively that solve

$$\{\mathbf{g}_u^{\text{opt}}, \mathbf{v}_u^{\text{opt}}\} = \arg \max_{\substack{\forall \mathbf{g}_u \in \mathcal{W} \\ \forall \mathbf{v}_u \in \mathcal{F}}} \|\mathbf{g}_u^H \mathbf{H}_u \mathbf{v}_u\|$$

MS u sets $\mathbf{w}_u = \mathbf{g}_u^{\text{opt}}$

BS sets $\mathbf{A} = [\mathbf{v}_1^{\text{opt}}, \mathbf{v}_2^{\text{opt}}, \dots, \mathbf{v}_U^{\text{opt}}]$

Second stage: Multuser digital precoding design

For each MS u , $u = 1, 2, \dots, U$

MS u estimates its effective channel $\bar{\mathbf{h}}_u = \mathbf{w}_u^H \mathbf{H}_u \mathbf{A}$

MS u feeds back $\bar{\mathbf{h}}_u$ to the BS

BS employs ZF digital precoder $\mathbf{D} = \bar{\mathbf{H}}^H (\bar{\mathbf{H}} \bar{\mathbf{H}}^H)^{-1}$, where $\bar{\mathbf{H}} = [\bar{\mathbf{h}}_1^T, \bar{\mathbf{h}}_2^T, \dots, \bar{\mathbf{h}}_U^T]^T$

$$\mathbf{d}_u = \frac{\mathbf{d}_u}{\|\mathbf{A} \mathbf{d}_u\|_F}, u = 1, 2, \dots, U$$

Multiuser Hybrid Precoding

- In first stage , using the beam training algorithms to design the analog precoding and combining vectors (without channel estimation)
- In second stage , each MS u estimates the effective channel $\bar{\mathbf{h}}_u = \mathbf{w}_u^H \mathbf{H}_u \mathbf{A}$. Then, $\bar{\mathbf{h}}_u$ is fed back to BS. ($\bar{\mathbf{h}}_{U \times 1} \square \mathbf{H}_u$)
- Finally , BS uses ZF digital precoder via $\bar{\mathbf{H}} = [\bar{\mathbf{h}}_1, \bar{\mathbf{h}}_2, \dots, \bar{\mathbf{h}}_U]$ to find \mathbf{D} . (with normalization)

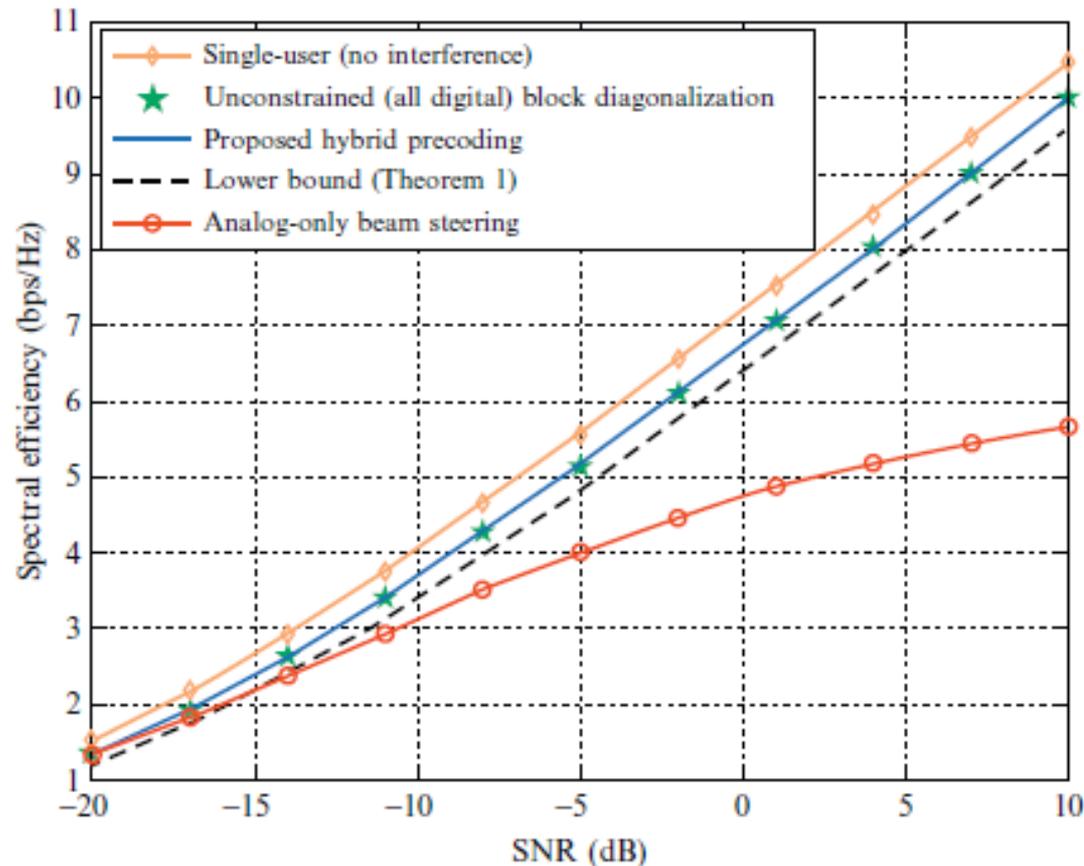
Multiuser Hybrid Precoding

<Performance Evaluation>

□ Simulation parameters:

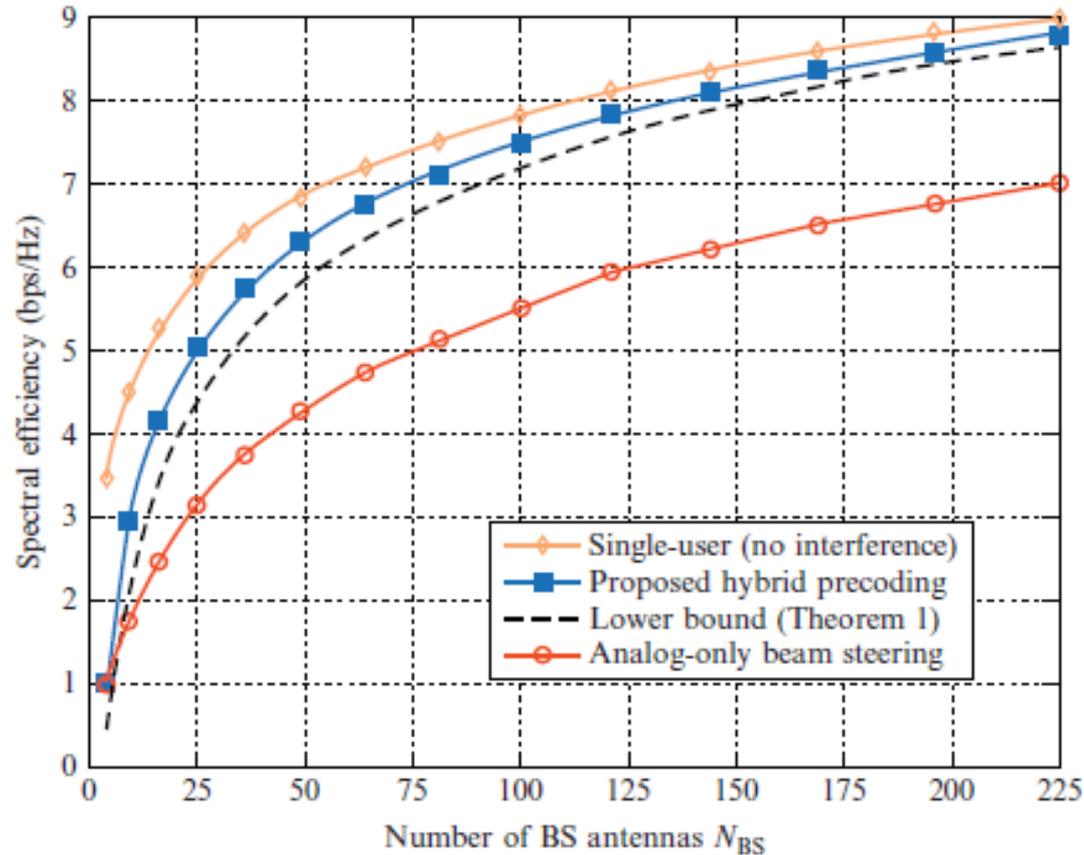
- BS: 8×8 UPA with 4 MSs , AZ AoAs/AoDs with uniformly $[0, 2\pi]$
- MS: each having 4×4 UPA. , EL AoAs/AoDs with uniformly $[-\pi/2, \pi/2]$
- Channel of each user has “ one path “

Multiuser Hybrid Precoding



Achievable rate comparison for an $N_{BS} = 64$, $N_{MS} = 16$, and $U = 4$ multiuser mmWave massive MIMO system.

Multiuser Hybrid Precoding



Achievable rate comparison against the number of BS antennas N_{BS} .

6. Conclusions

- ❑ Digital precoding aims to cancel data streams interference.
- ❑ Analog beamforming designs to improve the antenna array gain.
- ❑ Hybrid precoding combines the advantages of digital precoding and analog beamforming.
- ❑ Hybrid precoding seems more appropriate for future mmWave massive MIMO systems.
- ❑ Only a small-size digital precoder via a smaller number of RF chains will be required.



Thanks for your attention!