Learning-based Max-Min Fair Hybrid Precoding for mmWave Multicasting

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IEEE International Conference on Communications (ICC 2020) WC5: Machine Learning I (3rd Paper)



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 Multicast beamforming with fully-digital precoders has been widely studied in the literature.



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- Multicast beamforming with fully-digital precoders has been widely studied in the literature.
- However, the benefits and challenges with hybrid precoders require additional study.



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- Multicast beamforming with fully-digital precoders has been widely studied in the literature.
- However, the benefits and challenges with hybrid precoders require additional study.
- We investigate the joint design of hybrid precoding and analog combining for max-min fairness single-group multicasting in millimeter-wave systems. We propose LB-GDM, a learning-based approach that leverages (i) gradient descent with momentum and (ii) alternating optimization.

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- Features of the proposed scheme LB-GDM
 - Has low complexity [compared to SDR]
 - Leverages alternating optimization [several parameters]
 - Is based on learning with gradient descent with momentum
- Our proposed design does not require:
 - Code-books
 - Solution with a fully-digital precoder.

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System Model

Single-group Multicasting

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Hybrid Precoder

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Figure: Hybrid and fully-digital precoders

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System Model

The downlink signal is

$$\mathbf{x} = \mathbf{F}\mathbf{m}s\tag{1}$$

The received signal by user $k \in \mathcal{K}$ is

$$y_k = \underbrace{\mathbf{w}_k^H \mathbf{H}_k \mathbf{x}}_{\text{multicast signal}} + \underbrace{\mathbf{w}_k^H \mathbf{n}_k}_{\text{noise}}, \tag{2}$$

 \mathbf{w}_k : combiner of the k-th user \mathbf{F} : analog precoder \mathbf{m} : digital precoder \mathbf{H}_k : channel between the gNodeB and the k-th user K: number of users $\mathcal{K} = \{1, \dots, K\}$: set of users s: multicast symbol

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System Model

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The received signal by user $k \in \mathcal{K}$ is

$$y_k = \underbrace{\mathbf{w}_k^H \mathbf{H}_k \mathbf{Fms}}_{\text{multicast signal}} + \underbrace{\mathbf{w}_k^H \mathbf{n}_k}_{\text{noise}}, \tag{3}$$

The SNR at user k is

$$\gamma_k = \frac{\left|\mathbf{w}_k^H \mathbf{H}_k \mathbf{F} \mathbf{m}\right|^2}{\sigma^2 \left\|\mathbf{w}_k\right\|_2^2},\tag{4}$$

- \mathbf{w}_k : combiner of the k-th user
- \mathbf{F} : analog precoder
- m: digital precoder
- \mathbf{H}_k : channel between the gNodeB and the k-th user
- K: number of users

$$\mathcal{K} = \{1, \dots, K\}$$
: set of users

s: multicast symbol

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Problem Formulation

$$\mathcal{P}_{0}^{\mathsf{hyb}} : \max_{\mathbf{F}, \mathbf{m}, \{\mathbf{w}_{k}\}_{k=1}^{K}} \min_{k \in \mathcal{K}} \frac{\left|\mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{Fm}\right|^{2}}{\sigma^{2} \left\|\mathbf{w}_{k}\right\|_{2}^{2}}$$
(5a)
s.t.
$$\|\mathbf{Fm}\|_{2}^{2} = P_{\mathsf{tx}}^{\max},$$
(5b)
$$[\mathbf{F}]_{q,r} \in \mathcal{F}, q \in \mathcal{Q}, r \in \mathcal{R},$$
(5c)
$$\|\mathbf{w}_{k}\|_{2}^{2} = P_{\mathsf{rx}}^{\max}, k \in \mathcal{K},$$
(5d)
$$[\mathbf{w}_{k}]_{l} \in \mathcal{W}, l \in \mathcal{L}, \forall k \in \mathcal{K},$$
(5e)

$$\begin{split} \mathcal{F} &= \left\{ \sqrt{\delta_{\mathrm{tx}}}, \ldots, \sqrt{\delta_{\mathrm{tx}}} e^{j \frac{2\pi (L_{\mathrm{tx}}-1)}{L_{\mathrm{tx}}}} \right\}: \text{ allowed phase shifts at the precoder} \\ \mathcal{W} &= \left\{ \sqrt{\delta_{\mathrm{rx}}}, \ldots, \sqrt{\delta_{\mathrm{rx}}} e^{j \frac{2\pi (L_{\mathrm{rx}}-1)}{L_{\mathrm{rx}}}} \right\}: \text{ allowed phase shifts at the combiners} \\ L_{\mathrm{tx}}: \text{ number of phase shifts at the precoder} \\ L_{\mathrm{rx}}: \text{ number of phase shifts at the combiners} \end{split}$$

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Proposed Solution

$$\min \qquad \frac{\left|\mathbf{w}_{k}^{H}\mathbf{H}_{k}\mathbf{Fm}\right|^{2}}{(6a)}$$

$$\mathcal{P}_{1}^{\mathsf{hyb}} : \max_{\mathbf{F}} \min_{k \in \mathcal{K}} \qquad \frac{|\mathbf{w}_{k} \mathbf{\Pi}_{k} \mathbf{F} \mathbf{\Pi}|}{\sigma^{2} P_{\mathsf{rx}}^{\mathsf{max}}}$$
(6a)
s.t.
$$\|\mathbf{Fm}\|_{2}^{2} = P_{\mathsf{tx}}^{\mathsf{max}},$$
(6b)

$$\|\mathbf{Fm}\|_2^2 = P_{\mathrm{tx}}^{\mathrm{max}},\tag{6b}$$

$$\left[\mathbf{F}\right]_{q,r} \in \mathcal{F}, q \in \mathcal{Q}, r \in \mathcal{R}.$$
 (6c)

$$\mathcal{P}_{2}^{\mathsf{hyb}} : \max_{\mathbf{m}} \min_{k \in \mathcal{K}} \qquad \left| \mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{Fm} \right|^{2}$$
(7a)

$$\left\|\mathbf{Fm}\right\|_{2}^{2} = P_{\mathrm{tx}}^{\mathrm{max}}.$$
 (7b)

$$\mathcal{P}_{3}^{\mathsf{hyb}} : \max_{\{\mathbf{w}_{k}\}_{k=1}^{K}} \min_{k \in \mathcal{K}} \qquad \frac{\left|\mathbf{w}_{k}^{H}\mathbf{H}_{k}\mathbf{Fm}\right|^{2}}{\sigma^{2} \left\|\mathbf{w}_{k}\right\|_{2}^{2}} \qquad (8a)$$

s.t.
$$[\mathbf{w}_{k}]_{l} \in \mathcal{W}, l \in \mathcal{L}, \forall k \in \mathcal{K}. \qquad (8b)$$

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Optimization of the Analog Precoder ${f F}$

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$$\mathcal{P}_{1}^{\mathsf{hyb}} : \max_{\mathbf{F}} \min_{k \in \mathcal{K}} \qquad \frac{\left| \mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{Fm} \right|^{2}}{\sigma^{2} P_{\mathrm{rx}}^{\max}} \tag{9a}$$
s.t.
$$\| \mathbf{Fm} \|_{2}^{2} = P_{\mathrm{tx}}^{\max}, \tag{9b}$$

$$[\mathbf{F}]_{q,r} \in \mathcal{F}, q \in \mathcal{Q}, r \in \mathcal{R}. \tag{9c}$$

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We equivalently recast $\mathcal{P}_1^{\mathsf{hyb}}$ as $\overline{\mathcal{P}}_1^{\mathsf{hyb}}$

$$\overline{\mathcal{P}}_{1}^{\mathsf{hyb}} : \max_{\mathbf{F}} \min_{k \in \mathcal{K}} \qquad \frac{\mathbf{m}^{H} \mathbf{F}^{H} \mathbf{H}_{k}^{H} \mathbf{w}_{k} \mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{F} \mathbf{m}}{\mathbf{m}^{H} \mathbf{F}^{H} \mathbf{F} \mathbf{m}} \qquad (10a)$$
s.t.
$$[\mathbf{F}]_{q,r} \in \mathcal{F}, q \in \mathcal{Q}, r \in \mathcal{R}. \qquad (10b)$$

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Optimization of the Analog Precoder F

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Instead of approaching (10), we propose to solve the surrogate problem (11), which consists of a weighted sum of all $\tau_k^F = \frac{\mathbf{m}^H \mathbf{F}^H \mathbf{H}_k^H \mathbf{w}_k \mathbf{w}_k^H \mathbf{H}_k \mathbf{Fm}}{\mathbf{m}^H \mathbf{F}^H \mathbf{Fm}}, \text{ as shown in (11)}$

$$\widehat{\mathcal{P}}_{1}^{\mathsf{hyb}} : \max_{\mathbf{F}} \qquad \sum_{k=1}^{K} c_{k} \frac{\mathbf{m}^{H} \mathbf{F}^{H} \mathbf{H}_{k}^{H} \mathbf{w}_{k} \mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{Fm}}{\mathbf{m}^{H} \mathbf{F}^{H} \mathbf{Fm}} \qquad (11a)$$
s.t. $[\mathbf{F}]_{q,r} \in \mathcal{F}, q \in \mathcal{Q}, r \in \mathcal{R}, \qquad (11b)$

where $c_k \geq 0$ denotes the k-th weighting factor

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Optimization of the Analog Precoder F

Notice that

$$\tau_{k}^{F} \leq \lambda_{\max} \left(\left(\mathbf{F}^{H} \mathbf{F} \right)^{-1} \mathbf{F}^{H} \mathbf{H}_{k}^{H} \mathbf{w}_{k} \mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{F} \right)$$
$$= \underbrace{\mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{F} \left(\mathbf{F}^{H} \mathbf{F} \right)^{-1} \mathbf{F}^{H} \mathbf{H}_{k}^{H} \mathbf{w}_{k}}_{J_{k}^{F}}, \tag{12}$$

where $\lambda_{\max}(\cdot)$ extracts the maximum eigenvalue. Upon replacing τ_k^F in (11) by its upper bound J_k^F , the problem collapses to

$$\widetilde{\mathcal{P}}_{1}^{\mathsf{hyb}} : \max_{\mathbf{F}} \qquad \sum_{k=1}^{K} c_{k} \mathbf{w}_{k}^{H} \mathbf{H}_{k} \mathbf{F} \left(\mathbf{F}^{H} \mathbf{F} \right)^{-1} \mathbf{F}^{H} \mathbf{H}_{k}^{H} \mathbf{w}_{k}, \qquad (13a)$$

s.t. $[\mathbf{F}]_{q,r} \in \mathcal{F}, q \in \mathcal{Q}, r \in \mathcal{R}.$ $(13b)$

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Optimization of the Analog Precoder F

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Algorithm 1: Optimization of the analog precoder

Input: The precoders $\mathbf{F}^{(t-1)}$, $\mathbf{m}^{(t-1)}$ and receive combiners $\left\{\mathbf{w}_{k}^{(t-1)}\right\}_{k=1}^{K}$ **Output:** The analog precoder $\mathbf{F}^{(t)}$ **Execute:**

1: Calculate the weights $c_k^{(t)}, \forall k \in \mathcal{K}$.

2: Compute $\nabla J^F = \sum_{k=1}^{K} c_k^{(t)} \nabla_{\mathbf{F}} J_k^F / \left\| \nabla_{\mathbf{F}} J_k^F \right\|_{\mathbf{F}}$.

3: Compute the normalized gradient $\nabla \tilde{J}_{F}^{(t)} = \nabla J^{F} / \left\| \nabla J^{F} \right\|_{\mathrm{F}}$.

4: Compute $\mathbf{F}^{(t)} = \mathbf{F}^{(t-1)} + \rho_F \mathbf{F}^{(t-1)}_{\text{best}} + \alpha_F \nabla \tilde{J}_F^{(t)}$.

5: Project
$$\left[\mathbf{F}^{(t)}\right]_{q,r} \leftarrow \Pi_{\mathcal{F}} \left[\mathbf{F}^{(t)}\right]_{q,r}$$
 onto \mathcal{F} to satisfy (8b).

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Optimization of the Digital Precoder ${f m}$

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Algorithm 2: Optimization of the digital precoder

Input: The precoders $\mathbf{F}^{(t)}$, $\mathbf{m}^{(t-1)}$ and receive combiners $\left\{\mathbf{w}_{k}^{(t-1)}\right\}_{k=1}^{K}$ **Output:** The digital precoder $\mathbf{m}^{(t)}$

Execute:

1: Calculate the weights
$$d_k^{(t)}, \forall k \in \mathcal{K}$$
.

2: Compute
$$\nabla J^M = \sum_{k=1}^K d_k^{(t)} \nabla_{\mathbf{m}} J_k^M / \left\| \nabla_{\mathbf{m}} J_k^M \right\|_2$$
.

3: Compute the normalized gradient
$$\nabla \tilde{J}_M^{(t)} = \nabla J^M / \left\| \nabla J^M \right\|_2$$
.

4: Compute
$$\mathbf{m}^{(t)} = \mathbf{m}^{(t-1)} + \rho_M \mathbf{m}_{\text{pest}}^{(t-1)} + \alpha_M \nabla \tilde{J}_M^{(t)}$$
.

5: Normalize
$$\mathbf{m}^{(t)} \leftarrow \sqrt{P_{\mathrm{tx}}^{\mathrm{max}}} \mathbf{m}^{(t)} / \left\| \mathbf{F} \mathbf{m}^{(t)} \right\|_{2}$$
.

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Optimization of the Analog Combiner \mathbf{w}_k

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Algorithm 3: Optimization of the *k*-th combiner

Input: The precoders $\mathbf{F}^{(t)}$, $\mathbf{m}^{(t)}$ and the receive combiner $\mathbf{w}_{k}^{(t-1)}$ Output: The receive combiner $\mathbf{w}_{k}^{(t)}$ Execute: 1: Compute $\nabla_{\mathbf{w}_{k}} J_{k}^{W}$.

2: Compute
$$\nabla_{\mathbf{w}_k} J_W^{(t)} = \nabla_{\mathbf{w}_k} J_k^{(t)} / \|\nabla_{\mathbf{w}_k} J_k^{(t)}\|_2$$
.
3: Compute $\mathbf{w}_k^{(t)} = \mathbf{w}_k^{(t-1)} + \rho_W \mathbf{w}_{\text{best }k}^{(t-1)} + \alpha_W \nabla_{\mathbf{w}_k} \tilde{J}_W^{(t)}$.

4: Project
$$\begin{bmatrix} \mathbf{w}_{k}^{(t)} \end{bmatrix}_{l} \leftarrow \Pi_{\mathcal{W}} \begin{bmatrix} \mathbf{w}_{k}^{(t)} \end{bmatrix}_{l}$$
 onto $\mathcal{W}, \forall l \in \mathcal{L}$ to satisfy (12b).

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Optimization Algorithm

$\begin{array}{llllllllllllllllllllllllllllllllllll$	
1: Assign $\left[\mathbf{F}^{(0)}\right]_{q,r} \leftarrow \delta, q = \{1, \dots, N_{\mathrm{tx}}\}, r \leftarrow \mod\left(q, N_{\mathrm{tx}}^{\mathrm{RF}}\right) + 1,$	
$\mathbf{m}^{(0)} \leftarrow \begin{bmatrix} 1 \ 0_{1 \times (N_{\text{tw}}^{\text{RF}} - 1)} \end{bmatrix}^T, \mathbf{w}_k^{(0)} \leftarrow \begin{bmatrix} 1 \ 0_{1 \times (N_{\text{tx}} - 1)} \end{bmatrix}^T, \forall k \in \mathcal{K}.$	
2: Assign $\mathbf{F}_{\text{best}} \leftarrow 0$, $\mathbf{m}_{\text{best}} \leftarrow 0$ and $\{\mathbf{w}_{\text{best},k}\} \leftarrow 0$.	
3: Assign $\alpha_F \leftarrow \alpha_{F_0}, \alpha_M \leftarrow \alpha_{M_0}, \alpha_W \leftarrow \alpha_{W_0}$.	
4: Assign $t \leftarrow 0, \gamma_T \leftarrow 0$.	
Execute: $\sum_{i=1}^{N} \int d\mathbf{r} (exploration phase)$	
6. for $i_{xyy} = 1, \dots, N_{xyy}$, do (exploration phase)	
7: Compute $\mathbf{F}^{(t)}$, $\mathbf{m}^{(t)}$, $\left\{\mathbf{w}_{k}^{(t)}\right\}_{k=1}^{K}$ via Algorithms 1, 2, 3.	
8: Find the minimum SNR, γ_{min} , among all users.	
9: if $\gamma_{\min} \geq \gamma_T$	
10: Assign $\mathbf{F}_{opt} \leftarrow \mathbf{F}^{(t)}, \mathbf{m}_{opt} \leftarrow \mathbf{m}^{(t)}, \{\mathbf{w}_{opt,k}\}_{k=1}^{K} \leftarrow \{\mathbf{w}_{k}^{(t)}\}_{k=1}^{K}$.	
11: Assign $\gamma_T \leftarrow \gamma_{\min}$.	
12: end if	
13: Update $\alpha_F \leftarrow 0.98 \ \alpha_F, \ \alpha_M \leftarrow 0.98 \ \alpha_M, \ \alpha_W \leftarrow 0.98 \ \alpha_W.$	
14: Increment $t \leftarrow t + 1$.	
15: end for	
16: Assign $\mathbf{F}_{\text{best}}^{(t)} \leftarrow \mathbf{F}_{\text{opt}}, \mathbf{m}_{\text{best}}^{(t)} \leftarrow \mathbf{m}_{\text{opt}}, \left\{ \mathbf{w}_{\text{best},k}^{(t)} \right\}_{k=1}^{K} \leftarrow \left\{ \mathbf{w}_{\text{opt},k} \right\}_{k=1}^{K}$.	
17: Randomize $\mathbf{F}^{(t)}$, $\mathbf{m}^{(t)}$ and $\left\{\mathbf{w}_{k}^{(t)}\right\}_{k=1}^{K}$ enforcing (3b) - (3f).	
18: Assign $\alpha_F \leftarrow \alpha_{F_0}, \alpha_M \leftarrow \alpha_{M_0}, \alpha_W \leftarrow \alpha_{W_0}$.	
19: end for	
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Simulation Results - Scenario I

Goal: Evaluate the impact of exploration $(N_{\rm xpr})$ and exploitation $(N_{\rm xpt})$

Description	Symbol	Value	Units
Transmit power	P_{tx}^{max}	30	dBm
Receive power	P _{rx} ^{max}	10	dBm
Noise power	σ^2	30	dBm
Number of users	K	30	-
Number of transmit antennas	N_{tx}	15	-
Number of receive antennas	N_{rx}	2	-
Number of RF chains (at the hybrid precoder)	N_{tx}^{RF}	6	-
Number of phase shifts at the precoder	L_{tx}	8	-
Number of phase shifts at the combiner	L_{rx}	4	-
Number of exploration instances	N_{xpr}	100	-
Number of exploitation instances	$L_{\rm xpt}$	100	-
Momentum factor	$\rho_F = \rho_M = \rho_W$	0.90	-
Diminishing learning factor	$\alpha_F=\alpha_M=\alpha_W$	0.98	-

Table: Simulation parameters



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Image: A mathematical states and a mathem

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Simulation Results - Scenario I

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Figure: Impact of exploration (N_{xpr}) and exploitation (N_{xpt}) .



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Simulation Results - Scenario II

Goal: Evaluate the impact of the number of antennas $N_{\rm tx}$ and $N_{\rm rx}$

Description	Symbol	Value	Units
Transmit power	P_{tx}^{max}	30	dBm
Receive power	$P_{\rm rx}^{\rm max}$	10	dBm
Noise power	σ^2	30	dBm
Number of users	K	50	-
Number of transmit antennas	N_{tx}	$\{8, 12, 16\}$	-
Number of receive antennas	N_{rx}	$\{1, 2, 3, 4, 5\}$	-
Number of RF chains (at the hybrid precoder)	N_{tx}^{RF}	2	-
Number of phase shifts at the precoder	L_{tx}	8	-
Number of phase shifts at the combiner	L_{rx}	4	-
Number of exploration instances	N_{xpr}	100	-
Number of exploitation instances	L_{xpt}	100	-
Momentum factor	$\rho_F = \rho_M = \rho_W$	0.90	-
Diminishing learning factor	$\alpha_F=\alpha_M=\alpha_W$	0.98	-

Table: Simulation parameters



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Simulation Results - Scenario II

 $\left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 1 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 2 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 3 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 4 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \left[\mathsf{H} \right] | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = 5 \end{array} \right] \left[\left[\begin{array}{c} \mathsf{H} | N_{\mathrm{rx}} = S \end{array} \right] \left[\left[\left[\begin{array}[\\[\\$ Minimum SNR 150SE [bps/Hz] 100 50 \overline{N}_{tx} = 16 =

Figure: Performance evaluation of LB–GDM for varying $N_{\rm tx}$ and $N_{\rm rx}$ in fully-digital (D) and hybrid (H) precoders.



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Luis F. Abanto-Leon

Learning-based Max-Min Fair Hybrid Precoding for mmWave Multicasting

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Simulation Results - Scenario III

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Goal: Compare the performance with a SDR-based scheme

Table: Simulation parameters

Description	Symbol	Value	Units
Transmit power	P_{tx}^{\max}	30	dBm
Receive power	P_{yy}^{thax}	10	dBm
Noise power	σ^2	30	dBm
Number of users	K	$\{25, 50, 75, 100\}$	-
Number of transmit antennas	N_{tx}	20	-
Number of receive antennas	$N_{\rm rx}$	3	-
Number of RF chains (at the hybrid precoder)	N_{tr}^{RF}	6	-
Number of phase shifts at the precoder	L_{tx}^{tx}	8	-
Number of phase shifts at the combiner	$L_{\rm rx}$	4	-
Number of exploration instances	$N_{\rm xpr}$	120	-
Number of exploitation instances	$L_{\rm xpt}$	120	-
Momentum factor	$\rho_F = \rho_M = \rho_W$	0.90	-
Diminishing learning factor	$\alpha_F = \alpha_M = \alpha_W$	0.98	-
Number of randomizations (SDR-C)	N_{rand}	$\{1, 10, 50, 100, 500, 1000\}$	- 🏔

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Simulation Results - Scenario III

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- We investigated the design of fully-digital and hybrid precoders for single-group multicasting using a learning-based scheme, LB-GDM.
- Our proposed low-complexity LB-GDM uses only matrix multiplications / additions and low-dimensional matrix inversion operations.
- We compare the performance of precoders based on SDR-C and LB-GDM. The results show that LB-GDM attains substantial additional gain for both digital and hybrid precoders.
- We corroborate the importance of incorporating more receive antennas. We achieve 75.7% and 100% gains in terms of the minimum SNR by increasing the number of receive antennas from one to two.



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System

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This work has been funded by the Deutsche Forschungsgemeinschaft (DFG) within the B5G-Cell project as part of the SFB 1053 MAKI.



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