

Time-domain channel estimation for wideband millimeter wave systems with hybrid architecture

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MIMO and massive MIMO systems at mmWave

MIMO @ mmWave



Massive MIMO and mmWave



mmWave and massive MIMO are key ingredients of 5G

Large antenna arrays needed
at Tx and Rx

Leverage beamforming gain to
realize large data rates

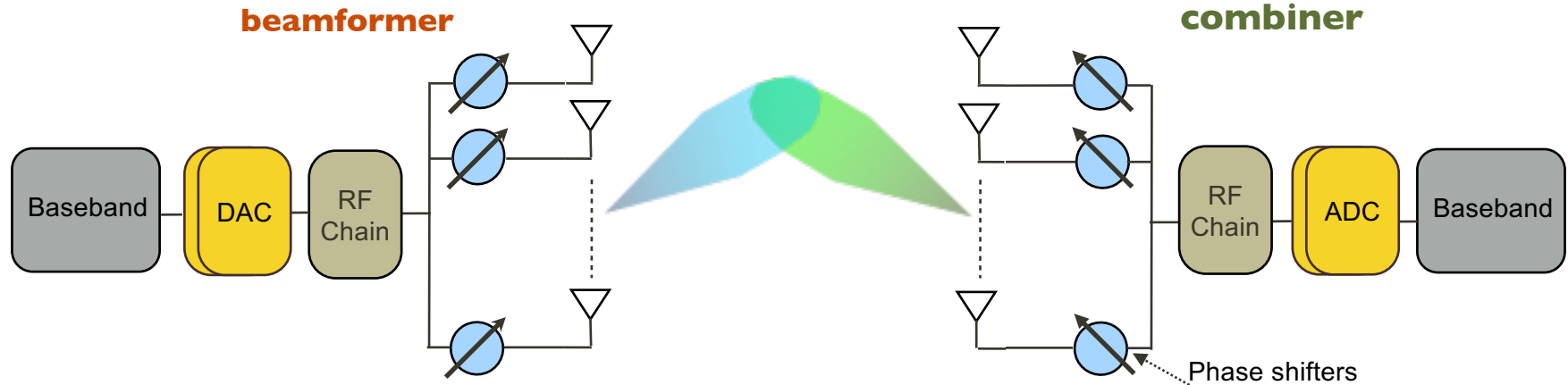
Large matrix sizes also
complicate channel estimation

Low-overhead channel estimation
techniques needed

Crucial for efficient precoder/combiner
design in both mmWave MIMO and
mmWave massive MIMO systems

MIMO channel estimation at mmWave is complicated due to hardware constraints

MIMO architectures at mmWave: analog beamforming



Phase shifters apply for the entire band

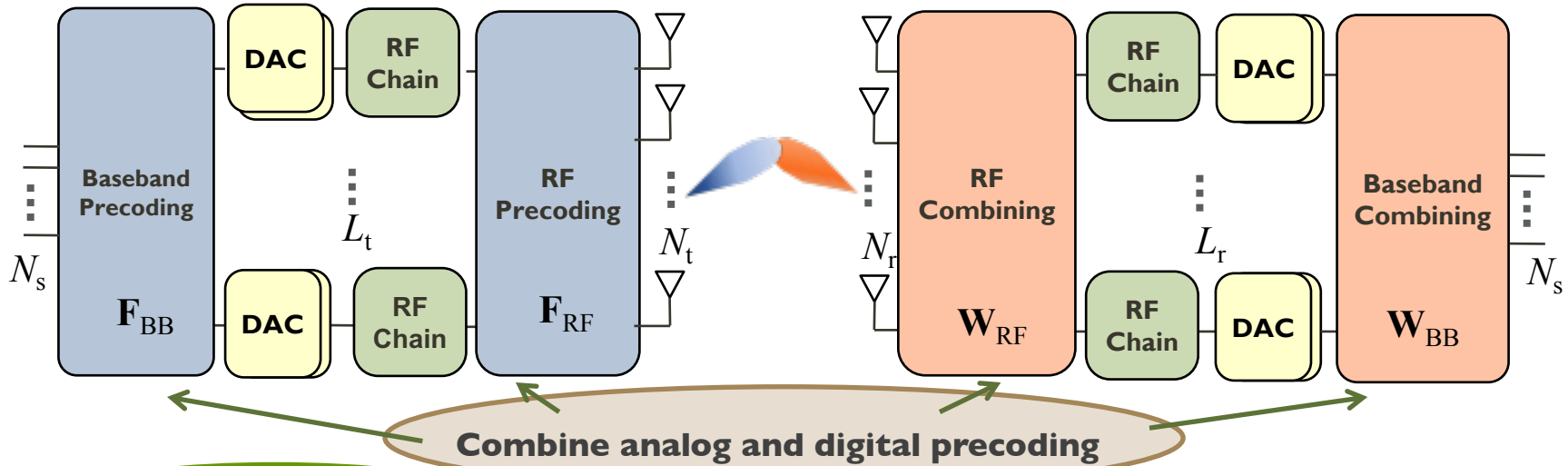
Beam training used to get the best Tx and Rx beamforming directions

Limited to single stream and single user MIMO

* J. Wang et al, "Beam codebook based beamforming protocol for multi-Gbps millimeter-wave WPAN systems," in *IEEE JSAC*, October 2009.

** S. Hur, T. Kim, D. Love, J. Krogmeier, T. Thomas, and A. Ghosh, "Millimeter wave beamforming for wireless backhaul and access in small cell networks," *IEEE Transactions on Communications*, vol. 61, no. 10, pp. 4391–4403, 2013.

MIMO architectures at mmWave: hybrid precoding



Fully digital MIMO is not feasible at mmWave

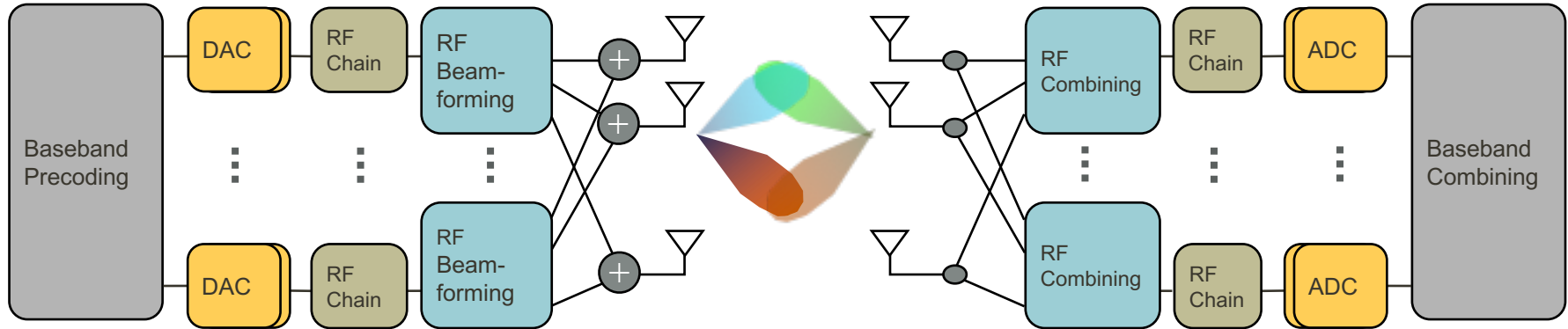
Rules out several digital channel estimation techniques

of DACs/ADCs \ll # of antennas

*Low resolution mixed circuit components not considered here

Hybrid precoding can support multi-stream and multi-user MIMO at mmWave

MmWave channel estimation with hybrid architecture



Multistream beam training →
large training overhead

Large bandwidth → frequency
selective channel

Low link SNR without
beamforming

No direct access to antenna outputs

Channel estimates are an alternative to beam training, work with multi-stream

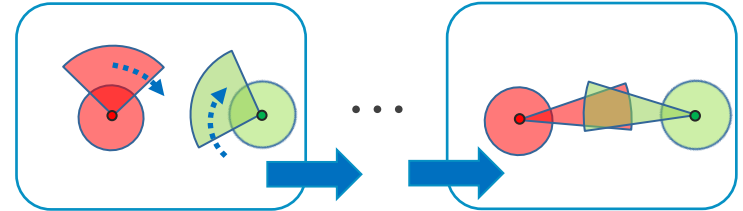
Prior work

Hierarchical beam training

Single stream support only

Used for analog architectures

Works for wideband channel



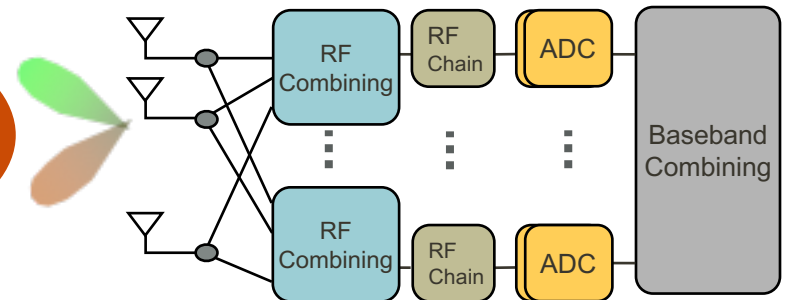
Avoids explicit channel estimation

Support for multi stream

Sparsity-based channel estimation

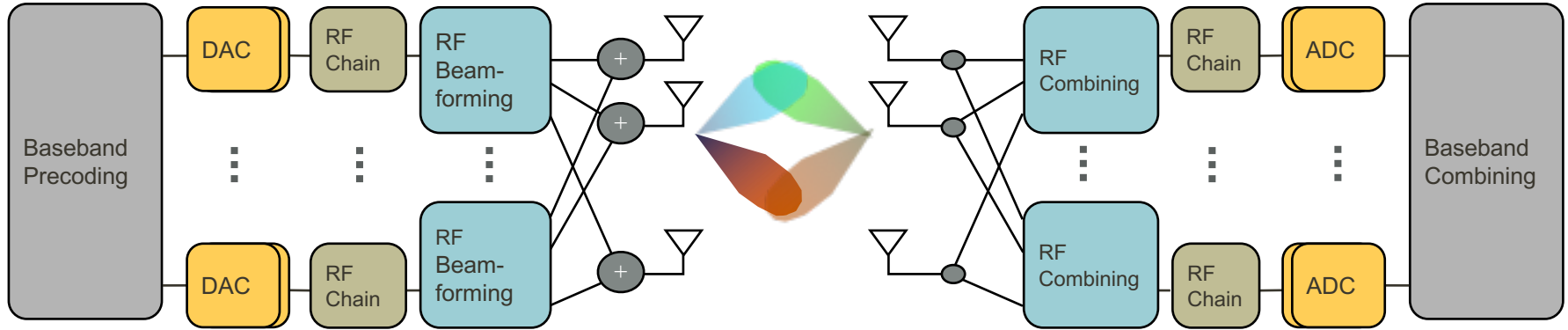
Wideband systems assumed ideal setting

Most prior work on narrowband channel model



Works for any architecture

Contributions



Wideband mmWave channel estimation technique

Sparse problem formulation in time domain

Useful in single-carrier mmWave systems

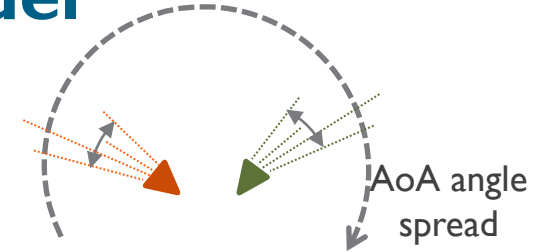
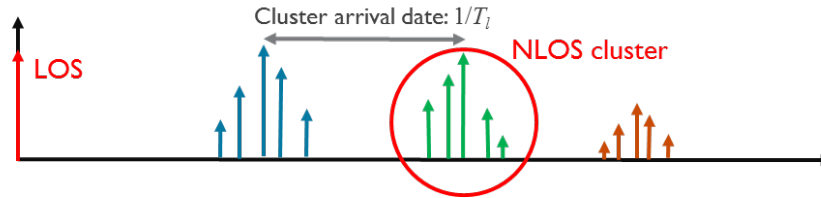
Include system constraints

- Frame structure
- Finite bandwidth of pulse shaper
- Hybrid architecture

Works for both MIMO and massive MIMO wideband mmWave systems

Wideband mmWave channel model

Clustered in time:



Clustered in space

$$L \ll N_r N_t$$

$$\mathbf{H}_d \in \mathbb{C}^{N_r \times N_t}$$

$$d = 0, 1, \dots, N_c - 1$$

of tap channel

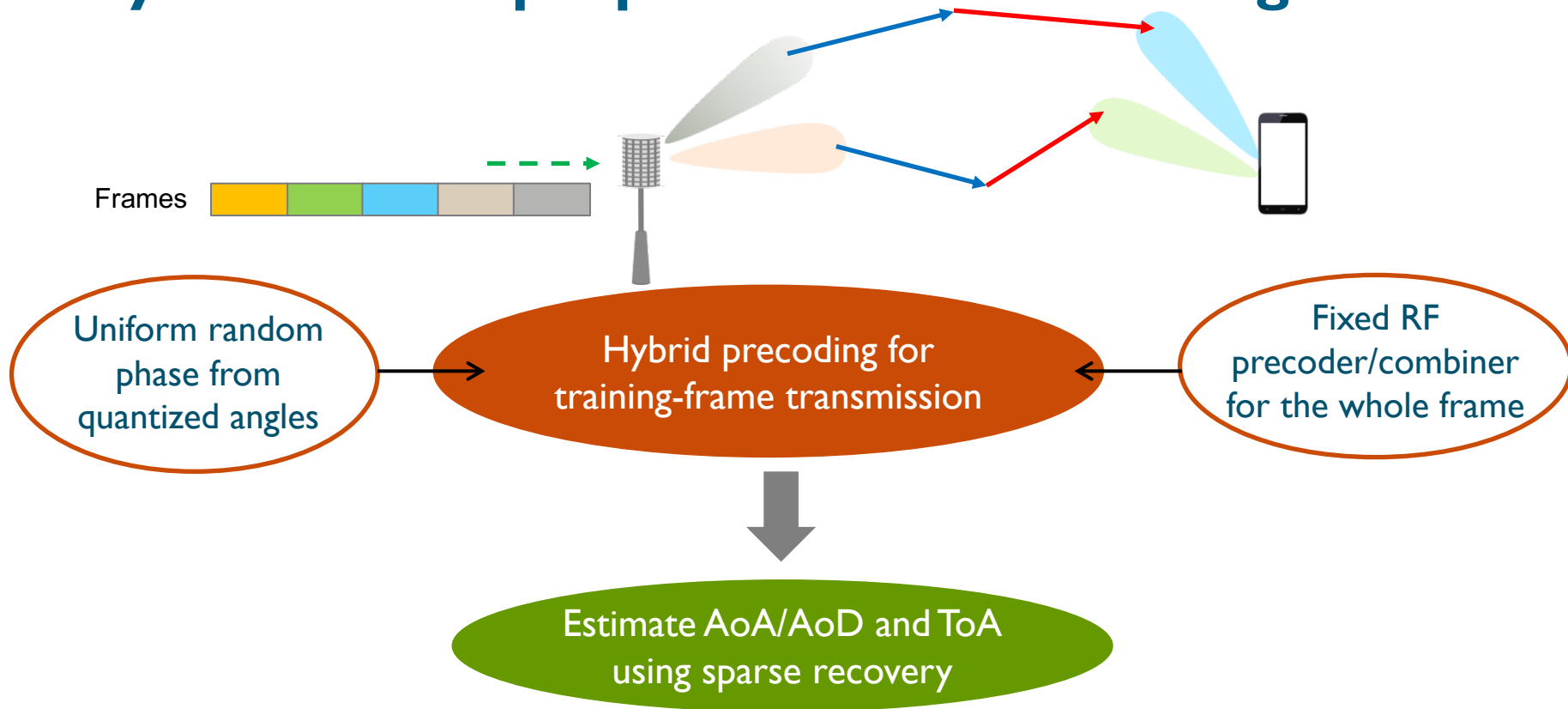
$$\mathbf{H}_d = \sum_{\ell=1}^L \alpha_{\ell} p_{\text{rc}}(dT_s - \tau_{\ell}) \mathbf{a}_R(\phi_{\ell}) \mathbf{a}_T^*(\theta_{\ell})$$

Complex path gain Pulse shaping function Path delay Angle of arrival/departure Antenna array response

Obtain $\{\phi_{\ell}, \theta_{\ell}, \alpha_{\ell}, \tau_{\ell}\}$ for channel estimation

Exploit sparsity in the angular and delay domain in the problem formulation

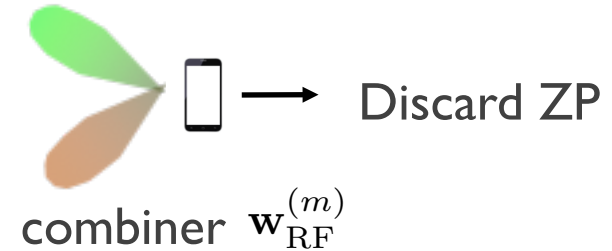
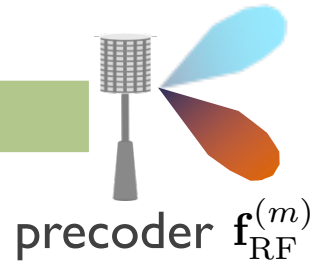
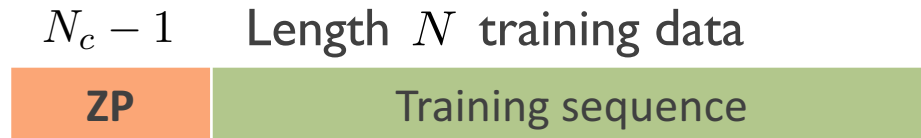
Key idea of the proposed channel training



Leverage the sparse structure in the mmWave channel & the hybrid architecture

Channel training stages

(zeros used for beam switching)

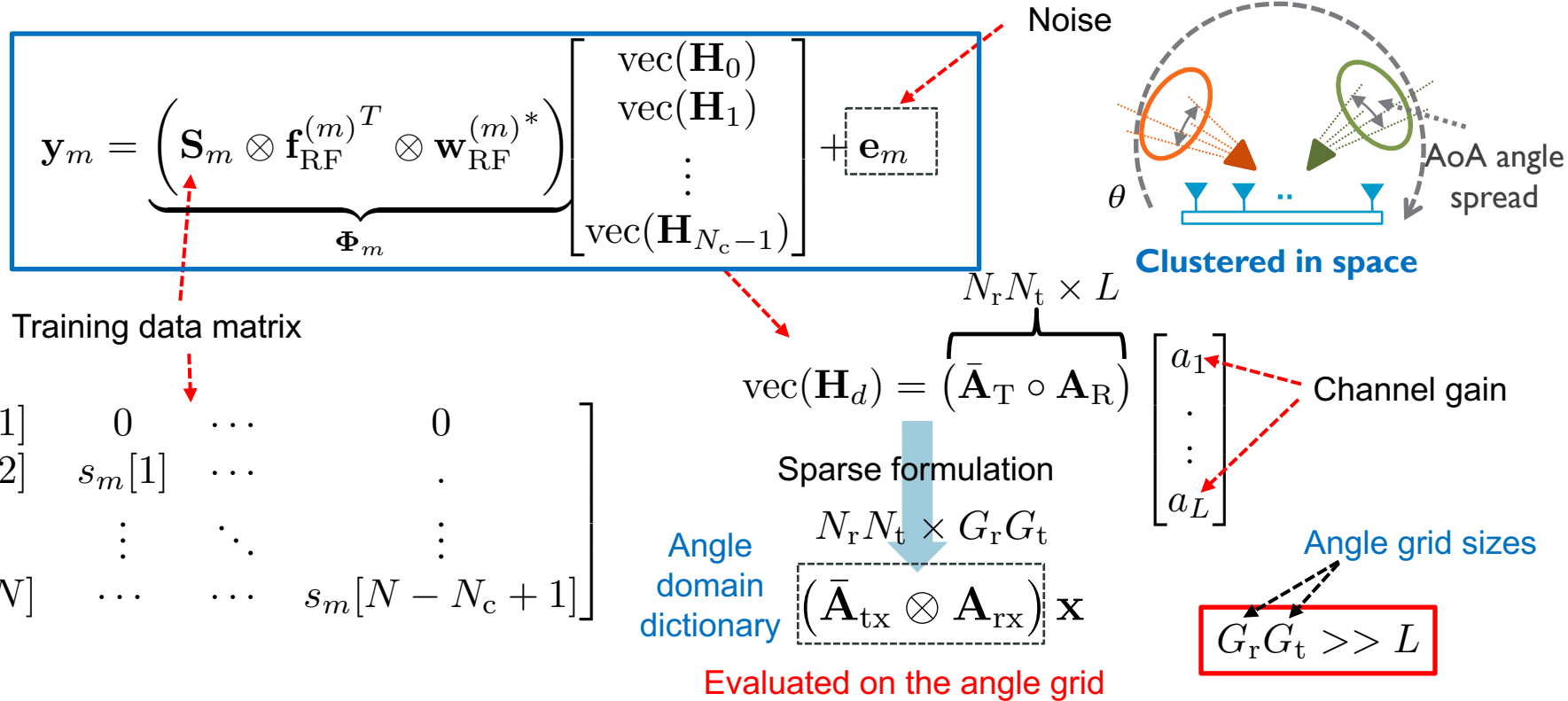


m^{th} training frame $\left[\underbrace{0 \cdots 0}_{N_c - 1} s_m[1] \cdots s_m[N] \right]$

$$\begin{bmatrix} y_m[1] \\ y_m[2] \\ \vdots \\ y_m[N] \end{bmatrix}^T = \mathbf{w}_{\text{RF}}^{(m)*} \begin{bmatrix} \mathbf{H}_0 & \cdots & \mathbf{H}_{N_c - 1} \end{bmatrix} \begin{bmatrix} \mathbf{f}_{\text{RF}}^{(m)} s_m[1] & \mathbf{f}_{\text{RF}}^{(m)} s_m[2] & \cdots & \mathbf{f}_{\text{RF}}^{(m)} s_m[N] \\ 0 & \mathbf{f}_{\text{RF}}^{(m)} s_m[1] & \cdots & \vdots \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 \cdots & \mathbf{f}_{\text{RF}}^{(m)} s_m[N - N_c + 1] \end{bmatrix} + \mathbf{e}^{(m)}$$

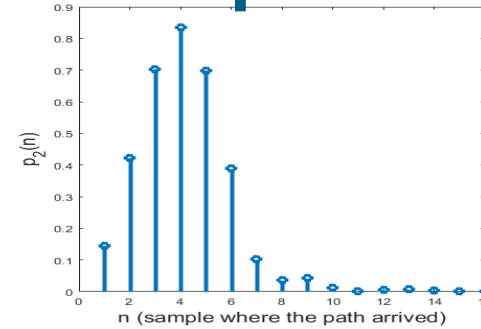
Zero-padding facilitates RF circuit reconfiguration across frames

Exploiting sparsity in the angular domain



Goal: Estimate the non-zeros elements of the sparse vector \mathbf{x}

Exploiting group sparsity due to pulse shaping



$$G_c \gg N_c$$

Delay grid sizes

$$p_d(n) = p_{rc} \left(\left(d - n \frac{N_c}{G_c} \right) T_s \right)$$

Pulse shaping function

Sampled version

$$\mathbf{p}_d \text{ has entries } p_d(n) \quad n = 1, 2, \dots, G_c$$

$$d = 0, 1, \dots, N_c - 1$$

$$\mathbf{y}_m = \Phi_m \left(\mathbf{I}_{N_c} \otimes \bar{\mathbf{A}}_{tx} \otimes \mathbf{A}_{rx} \right) \Gamma \mathbf{x} + \mathbf{e}_m,$$

where

Delay domain dictionary

$$\Gamma = \begin{bmatrix} \mathbf{I}_{G_r G_t} \otimes \mathbf{p}_0^T \\ \mathbf{I}_{G_r G_t} \otimes \mathbf{p}_1^T \\ \vdots \\ \mathbf{I}_{G_r G_t} \otimes \mathbf{p}_{N_c-1}^T \end{bmatrix}$$

Evaluated on the delay grid

Unknown \mathbf{x} is $G_t G_r G_c \times 1$, L -sparse vector containing the complex channel gains

Compressive channel estimation

Stack M measurements $\rightarrow \mathbf{y} = \sqrt{\rho} \Phi \Psi \mathbf{x} + \mathbf{e}$

Effective dictionary matrix

Measurement matrix

Measurement 1 $\mathbf{y}_1 = \sqrt{\rho} \left(\mathbf{S}_1 \otimes \mathbf{f}_{\text{RF}}^{(1)T} \otimes \mathbf{w}_{\text{RF}}^{(1)*} \right) \left(\mathbf{I}_{N_c} \otimes \bar{\mathbf{A}}_{\text{tx}} \otimes \mathbf{A}_{\text{rx}} \right) \Gamma \mathbf{x} + \mathbf{e}_1$

⋮

Measurement M $\mathbf{y}_M = \sqrt{\rho} \left(\mathbf{S}_M \otimes \mathbf{f}_{\text{RF}}^{(M)T} \otimes \mathbf{w}_{\text{RF}}^{(M)*} \right) \left(\mathbf{I}_{N_c} \otimes \bar{\mathbf{A}}_{\text{tx}} \otimes \mathbf{A}_{\text{rx}} \right) \Gamma \mathbf{x} + \mathbf{e}_M$

Contains quantized grid of ToA

Random beamforming matrices

Dictionary with columns

$$\mathbf{a}_{\text{T}}^c(\tilde{\phi}_x) \otimes \mathbf{a}_{\text{R}}(\tilde{\theta}_y)$$

Quantized grid of AoA/AoD

Dictionary matrix constructed using antenna array response

Angle grid & delay quantization can be made as fine as required for sparsity

Extends directly to multiple RF chains during training

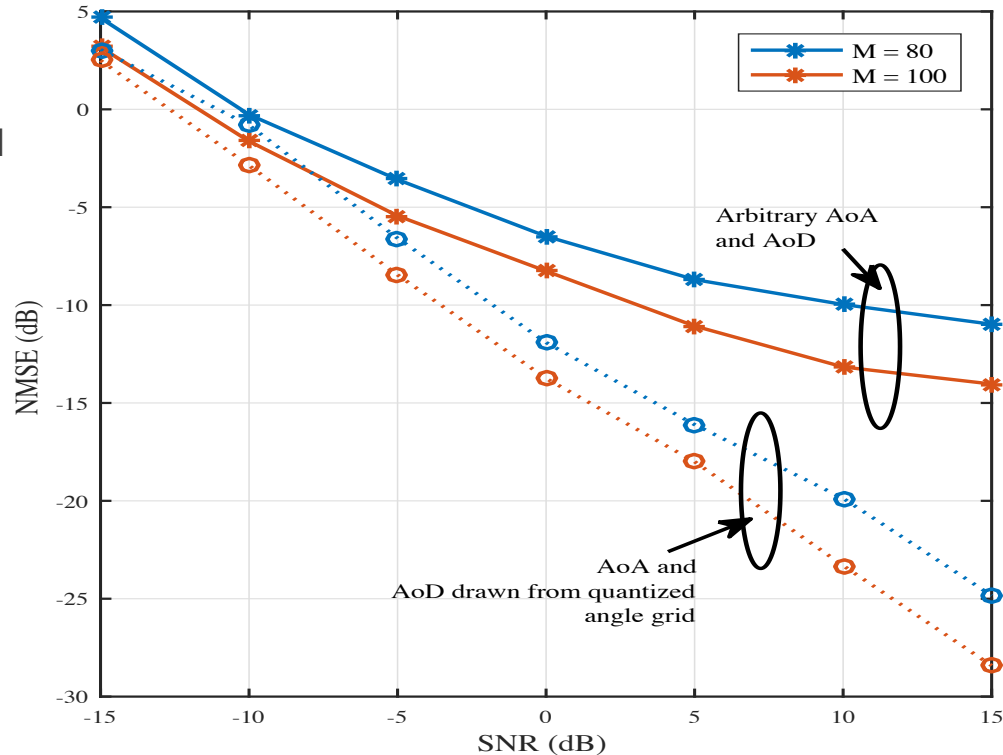
Simulation results

Setup

- Tx has 32 antennas, Rx has 32 antennas
- Dictionary generated using AoD/AoA with grid size = 64
- Frequency selective channel with 4 delay taps and 2 paths
- Pulse shaping filter with 0.8 roll-off factor
- Frame length = 16
- 2 bit quantization for precoder and combiner phase shifters
- **Orthogonal Matching pursuit** followed by **least square estimation**

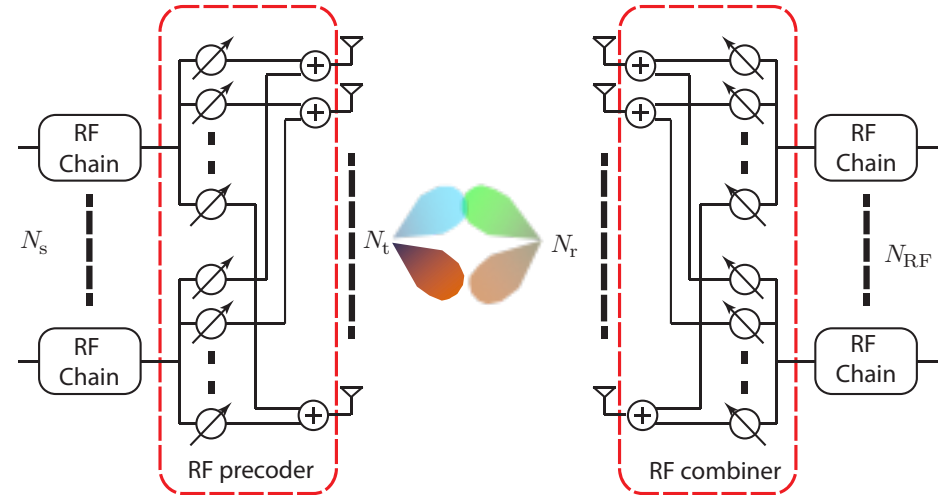
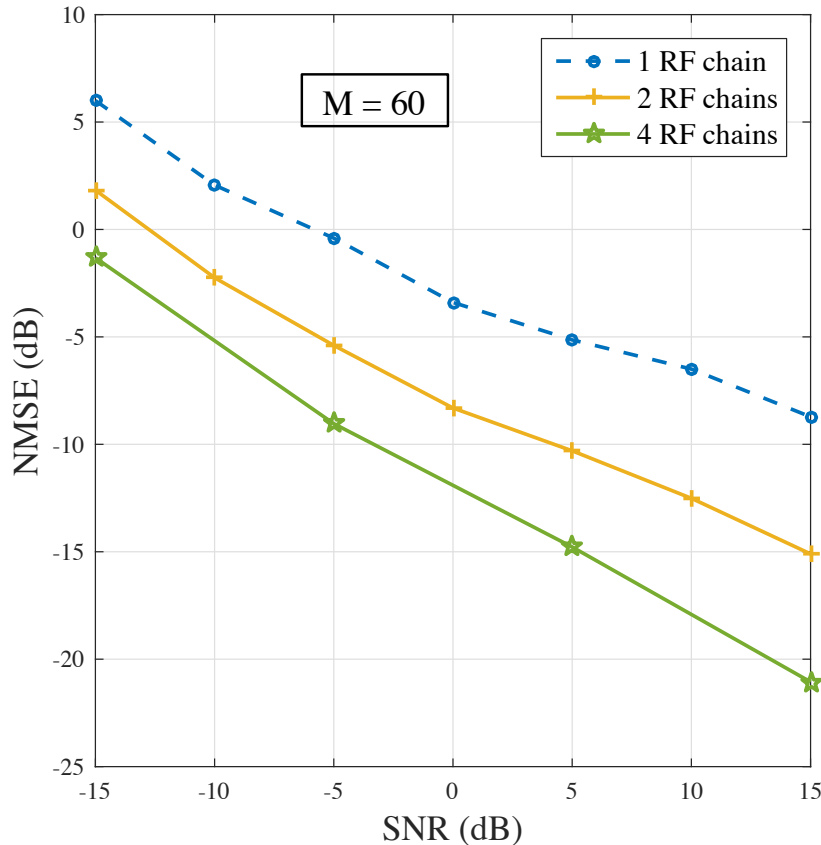
$$\text{NMSE} = \frac{\sum_{d=0}^{N_c} \|\mathbf{H}_d - \hat{\mathbf{H}}_d\|_F^2}{\sum_{d=0}^{N_c} \|\mathbf{H}_d\|_F^2}$$

With 1 RF chain



80-100 training frames are enough to ensure low channel estimation error

Employing hybrid architecture



Effectively increases the precoding and combining beam patterns

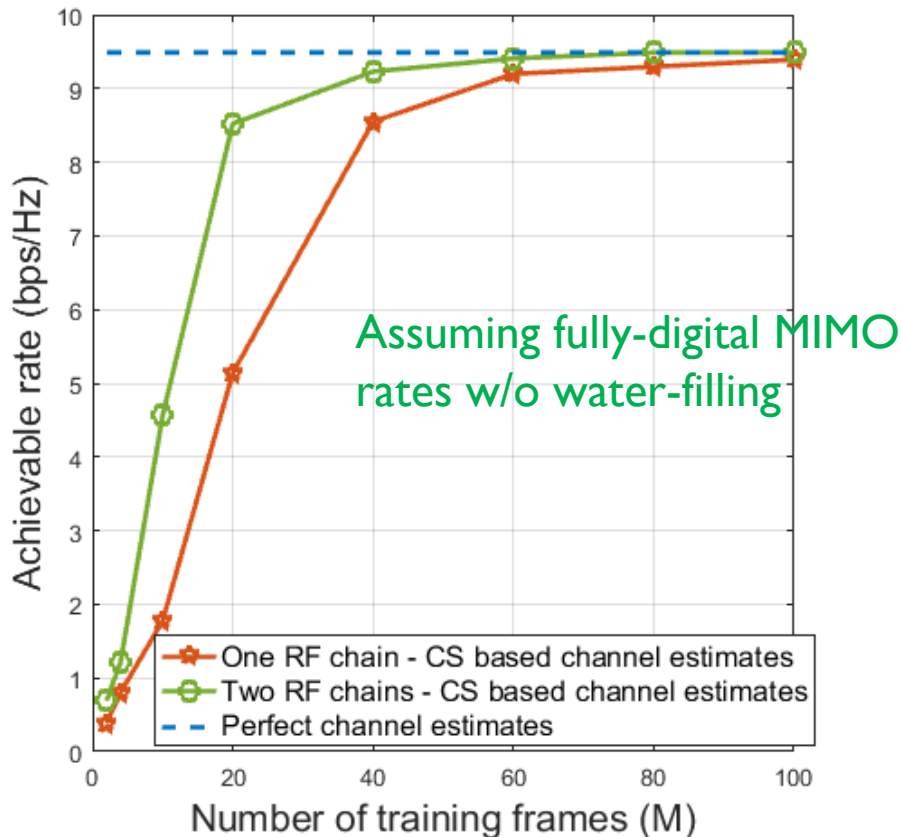
Using multiple RF chains at Tx and Rx gives better channel estimates

Reducing the training overhead

Leverage multiple RF chains

Increased number of measurements per training frame

Hybrid precoding can give rates close to fully-digital MIMO



Using multiple RF chains reduces training overhead

Conclusion and future work

Wideband mmWave channel estimation needs to consider **hardware constraints**

- ✦ Fewer number of baseband measurements
- ✦ Effective baseband channel is less sparse

Proposed time domain channel estimation using hybrid architecture

- ✦ Sparse formulation enables use of compressive sensing tools
- ✦ Multiple RF chains at the transceivers reduce the number of training step

Future work

- ✦ Compare complexity with frequency domain channel estimation techniques
- ✦ Comparison of performance between beam training and CS based approaches

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