Channel Estimation and Hybrid Precoding for Millimeter Wave Systems

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mmWave Cellular Systems

Current systems live with
 limited microwave spectrum



Potential large bandwidth available at high frequencies [I]



300 MHz 3 GHz 28 GHz 38 GHz 80 GHz

- 300 GHz
- mmWave band can enable multi-gigabit data rates
- Hardware technology advances make mmWave possible

[1] Z. Pi and F. Khan, "An introduction to millimeter-wave mobile broad- band systems," *IEEE Communications Magazine*, vol. 49, no. 6, pp. 101–107, 2011.

Directional Beamforming

- mmWave communication needs high beamforming gain <u>Why?</u>
- High frequency signals experience severe path-loss
- Larger bandwidth at mmWave means higher noise power Solution



Leverage array gain from large antenna arrays

Limitations of mmWave beamforming

- High cost of mixed-signal components limits RF chains
 <u>Possible solution</u>
- Perform the beamforming processing in analog (RF) domain



- But RF imposes other constraints that limits the performance Ex. Phase shifter may have constant gains and quantized angles
- Managing multi-stream and multi-user interference is difficult Need new precoding solutions for mmWave systems

mmWave-suitable Precoding



Hybrid Analog/digital Precoding Architecture

- Hybrid precoding (HP) Enables
 - Multiplexing multiple data streams
 - Manage interference between multiple users

Prior work considers designing HP with full channel knowledge [1]

But, is it simple to estimate mmWave channels?

[1] O. El Ayach, S. Rajagopal, S. Abu-Surra, Z. Pi, and R. Heath, "Spatially sparse precoding in millimeter wave MIMO systems," IEEE Transactions on Wireless Communications, vol. 13, no. 3, pp. 1499–1513, March 2014

Hybrid Precoding Design

The hybrid precoding matrices are jointly designed to solve[]

Minimize the Frobenius norm of the error between the optimal precoder (singular value decomposition) and the constrained hybrid precoders

$$\begin{aligned} (\mathbf{F}_{\mathrm{RF}}^{\star}, \mathbf{F}_{\mathrm{BB}}^{\star}) &= \arg\min & \|\mathbf{F}_{\mathrm{opt}} - \mathbf{F}_{\mathrm{RF}}\mathbf{F}_{\mathrm{BB}}\|_{F}, \\ \text{s.t.} & [\mathbf{F}_{\mathrm{RF}}]_{:,i} \in \left\{ [\mathbf{A}_{\mathrm{can}}]_{:,\ell} \mid 1 \leq \ell \leq N_{\mathrm{can}} \right\}, \\ & \|\mathbf{F}_{\mathrm{RF}}\mathbf{F}_{\mathrm{BB}}\|_{F}^{2} = N_{\mathrm{S}}. \end{aligned}$$

The set of quantized (candidate) array response vectors

- Main Idea: Approximate the optimal precoders
- This is a sparse approximation problem

Matching pursuit algorithms can be leveraged in the design

Matching Pursuit Based HP Design

Algorithm 1 Hybrid Analog-Digital Beamforming Design

$$\begin{split} \mathcal{R} &= \phi \\ \mathbf{F}_{\mathrm{res}} = \mathbf{F}_{\mathrm{opt}} \\ \text{for } i &\leq N_{\mathrm{RF}} \text{ do} \\ \mathbf{\Psi} &= \mathbf{A}_{\mathrm{can}}^{H} \mathbf{F}_{\mathrm{res}} \\ k &= \arg\max_{j=1,2,..N_{\mathrm{can}}} \left(\mathbf{\Psi}^{H} \mathbf{\Psi} \right)_{j,j}^{\mathrm{Find the direction from } \mathbf{A}_{\mathrm{can}}} \\ k &= \arg\max_{j=1,2,..N_{\mathrm{can}}} \left(\mathbf{\Psi}^{H} \mathbf{\Psi} \right)_{j,j}^{\mathrm{Find the direction from } \mathbf{A}_{\mathrm{can}}} \\ \mathbf{R} &= \mathcal{R} \cup k \\ \mathbf{F}_{\mathrm{RF}} &= \left[\mathbf{A}_{\mathrm{can}} \right]_{:,\mathcal{R}} \\ \mathbf{F}_{\mathrm{BB}} &= \left(\mathbf{F}_{\mathrm{RF}}^{H} \mathbf{F}_{\mathrm{RF}} \right)^{-1} \mathbf{F}_{\mathrm{RF}}^{H} \mathbf{F}_{\mathrm{opt}} \\ \mathbf{F}_{\mathrm{res}} &= \frac{\mathbf{F}_{\mathrm{res}} - \mathbf{F}_{\mathrm{RF}} \mathbf{F}_{\mathrm{BB}}}{\|\mathbf{F}_{\mathrm{res}} - \mathbf{F}_{\mathrm{RF}} \mathbf{F}_{\mathrm{BB}} \|_{F}} \end{split}$$

$$\begin{aligned} \mathsf{Project out the chosen \\ \mathsf{precoders from } \mathbf{F}_{\mathrm{res}} \\ \mathbf{F}_{\mathrm{BB}} &= \sqrt{N_{\mathrm{S}}} \frac{\mathbf{F}_{\mathrm{BB}}}{\|\mathbf{F}_{\mathrm{RF}} \mathbf{F}_{\mathrm{BB}}\|_{F}} \end{aligned}$$

mmWave Channel Estimation

mmWave channel estimation is challenging

- Large transmit and receive antenna arrays
- Low signal-to-noise ratio before beamforming
- Limited number of RF chains imposes constraints on training signal design

Prior work avoids explicit mmWave channel estimation

Idea: Beam training by searching for the best beamforming directions [1],[2]

Limitations of beam training

- It supports only single-stream transmission
- It requires high training overhead to search for the best beams

Note. Some ideas considered hierarchical search to minimize the training overhead.

It lacks the ability to manage multi-user transmissions

Need to design new mmWave channel estimation algorithms

[1] J. Wang, Z. Lan, C. Pyo, T. Baykas, C. Sum, M. Rahman, J. Gao, R. Funada, F. Kojima, H. Harada *et al.,* "Beam codebook based beamforming protocol for multi-Gbps millimeter-wave WPAN systems," *IEEE Journal on Selected Areas in Communications,* vol. 27, no. 8, pp. 1390–1399, 2009.

[2] 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.



Consider hybrid precoding architecture





Ns: Number of streams N_{RF}: Number of RF chains N_{BS}: Number of BS antennas N_{MS}: Number of MS antennas

Phase shifters have constant modulus and quantized angles

 $\mathbf{W} = \mathbf{W}_{\mathrm{RF}}\mathbf{W}_{\mathrm{BB}}$ $\mathbf{F} = \mathbf{F}_{\mathrm{RF}}\mathbf{F}_{\mathrm{BB}}$

In RF chains are much less than the number of antennas

mmWave Channels

We adopt a geometric channel model



Assumption

- Channels are sparse, a few paths exist [1]
- Arrays are generally non-uniform
- One of the channel paths is LOS

[1] M. Riza Akdeniz, Y. Liu, S. Sun, S. Rangan, T. S. Rappaport, and E. Erkip, "Millimeter Wave Channel Modeling and Cellular Capacity Evaluation," Dec. 2013. [Online]. Available: http://arxiv.org/abs/1312.4921



The derived formulation captures the channel sparsity

$$\mathbf{y}_{v} = \sqrt{P} \left(\mathbf{F}^{T} \otimes \mathbf{W}^{H} \right) \mathbf{A}_{D} \mathbf{z} + \mathbf{n}_{Q}$$

Dictionary of steering vectors. Its columns of the form $\left(\mathbf{a}_{\mathrm{BS}}^{*}\left(\bar{\phi}_{u}\right)\otimes\mathbf{a}_{\mathrm{MS}}\left(\bar{\theta}_{v}\right)\right)$

Sparse vector L non-zeros

- Advantages of the sparse formulation
 - Enables using tools from (adaptive) compressed sensing
 - Enables estimating multi-path channels
- We will adaptively design training precoders/combiners

Hierarchical Codebook Design

Construct multi-resolution beamforming codebooks



Codebook level 1 Codebook level 2 Codebook level 3 These patterns are generated with unconstrained (digital-only) precoders Approximation using hybrid analog/digital precoding



Each beam pattern is approximated using hybrid analog/digital architecture.

Simulation Results



- Achieve comparable performance to exhaustive search solutions
- Ifficiently estimate multipath channels
- Quantized angles assumption has a small impact on performance

*Note: the numbers of the RF chains are chosen just to be have reasonable values.

Conclusion

- Ochannel sparsity enables low-complexity hybrid precoding designs
- Exploiting channel sparsity, we developed efficient channel training and estimation algorithms
- The developed algorithm
 - estimates multipath channels
 - is more robust to angles quantization compared with analog-only solutions
 - considers practical mmWave hardware (hybrid precoding architecture)
- The performance of the proposed algorithm approaches that of exhaustive search

Future work:

- Extend the proposed algorithm to multi-user mmWave systems
- Consider other training beamforming designs (random beamforming)

Thank you