



# Tree-Structured Quantization on Grassmann and Stiefel Manifolds

*Data Compression Conference (virtual) 2021*

**Stefan Schwarz** and Markus Rupp  
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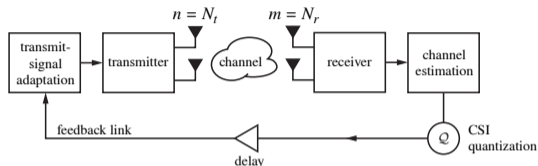
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# Motivation for Grassmann/Stiefel-Manifold Quantization



- Limited feedback MIMO single/multi-user wireless communications  $\mathbf{H} \in \mathbb{C}^{n \times m}$ ,  $m < n$

$$\mathbf{H} = \mathbf{U}\Sigma\mathbf{V}^H, \quad \mathbf{U}^H\mathbf{U} = \mathbf{I}_m$$

- Grassmannian feedback: transmit-subspace  $\text{span}(\mathbf{H}) = \text{span}(\mathbf{U})$
- Stiefel-manifold feedback: individual singular-vectors  $\mathbf{U}$

## Motivation for Grassmann/Stiefel-Manifold Quantization (II)

- Common approach for channels without structure (e.g., i.i.d. Rayleigh fading):

*Single-stage codebook-based quantization*

$$\mathbf{Q}^* = \mathcal{Q}(\mathbf{H}) = \arg \min_{\mathbf{Q}_\ell \in \mathcal{Q}_m^{(n)}} d(\mathbf{H}, \mathbf{Q}_\ell)$$

- Grassmannian and Stiefel-manifold quantization metrics

$$d(\mathbf{H}, \mathbf{Q}_\ell) = d_c^2(\mathbf{U}, \mathbf{Q}_\ell) = m - \text{tr}(\mathbf{U}^H \mathbf{Q}_\ell \mathbf{Q}_\ell^H \mathbf{U}),$$

$$d(\mathbf{H}, \mathbf{Q}_\ell) = \|\mathbf{U} - \mathbf{Q}_\ell\|_F^2 = 2m - 2\Re(\text{tr}(\mathbf{U}^H \mathbf{Q}_\ell))$$

- Single-stage quantization codebooks

$$\mathcal{Q}_m^{(n)} = \left\{ \mathbf{Q}_\ell \in \mathbb{C}^{n \times m} \mid \mathbf{Q}_\ell^H \mathbf{Q}_\ell = \mathbf{I}_m, \ell \in \{1, \dots, K\} \right\}$$

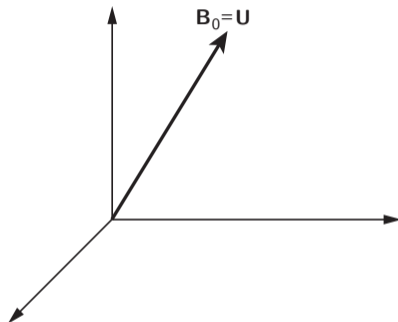
- Main issue: codebook size has to grow exponentially with dimensions  $n, m \Rightarrow$  complexity explodes

## **Multi-Stage Manifold Quantization**

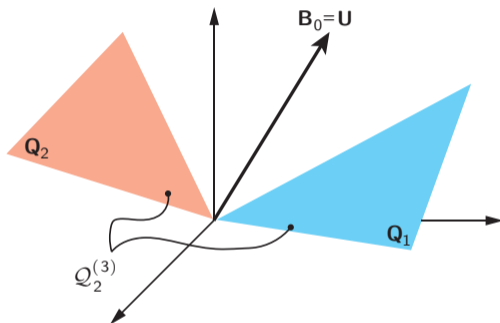
Performance Evaluation

Conclusions and Future Work

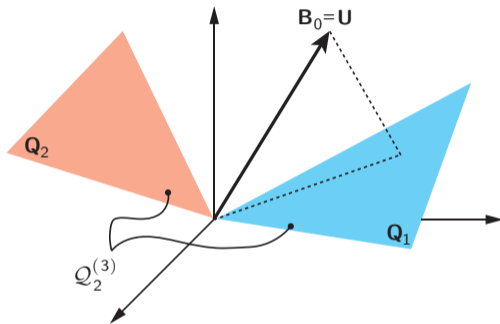
## Recursive (Greedy) Multi-Stage Grassmannian Quantization – Toy Example $n \times m = 3 \times 1$



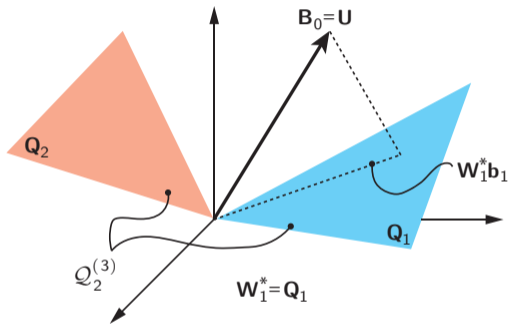
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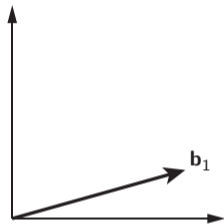
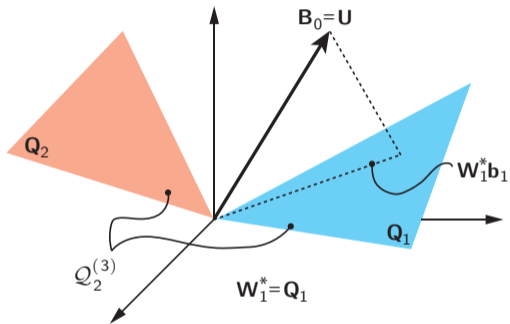


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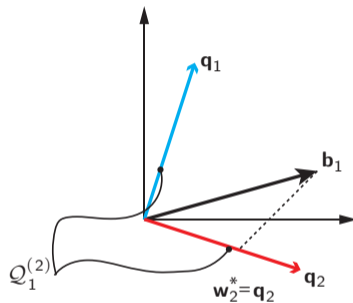
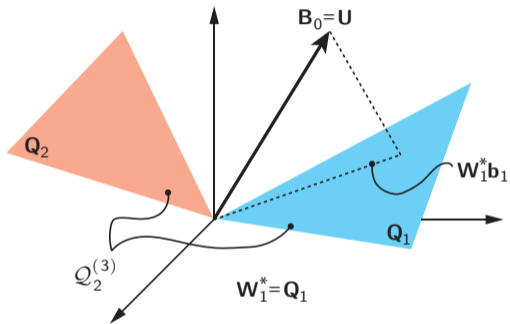




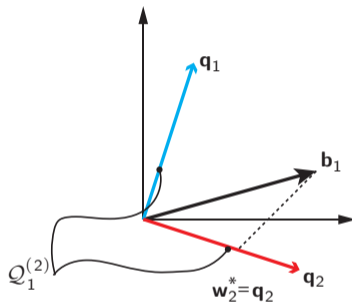
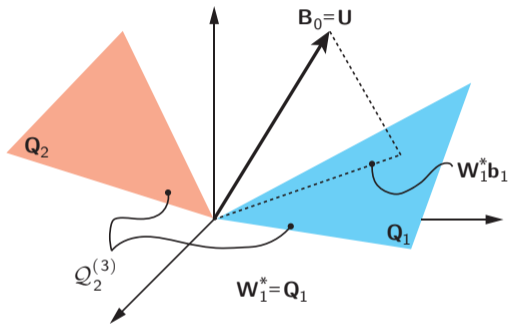
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$$Q^* = W_1^* w_2^*$$

## Recursive (Greedy) Multi-Stage Grassmannian Quantization

- Denote the initial quantizer input as  $\mathbf{B}_0 = \mathbf{U} \in \mathbb{C}^{n \times m}$  with  $d_0 = n$
- Recursive (greedy) multi-stage quantization in  $R$  stages  $i \in \{1, \dots, R\}$
- Input of stage  $i$ :  $\mathbf{B}_{i-1} \in \mathbb{C}^{d_{i-1} \times m}$
- Output of stage  $i$ : quantized subspace  $\mathbf{W}_i^* \in \mathbb{C}^{d_{i-1} \times d_i}$  and combining matrix  $\mathbf{B}_i \in \mathbb{C}^{d_i \times m}$
- Dimension shrinking:  $d_{i-1} > d_i$ ,  $d_0 = n$ ,  $d_R = m$

$$\mathbf{W}_i^* = \arg \min_{\mathbf{Q} \in \mathcal{Q}_{d_i}^{(d_{i-1})}} d_c^2(\mathbf{B}_{i-1}, \mathbf{Q}),$$

$$\mathbf{B}_i = \mathbf{W}_i^{*H} \mathbf{B}_{i-1} \left( \mathbf{B}_{i-1}^H \mathbf{W}_i^* \mathbf{W}_i^{*H} \mathbf{B}_{i-1} \right)^{-\frac{1}{2}}$$

- Subspace quantization-based combining (SQBC) matrix  $\mathbf{B}_i$ :  $d_c^2(\mathbf{B}_{i-1}, \mathbf{W}_i^*) = d_c^2(\mathbf{B}_{i-1}, \mathbf{W}_i^* \mathbf{B}_i)$
- Final quantizer output:  $\mathbf{Q}^* = \prod_{i=1}^R \mathbf{W}_i^* \in \mathbb{C}^{n \times m}$

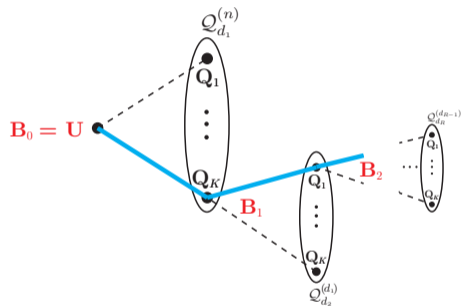
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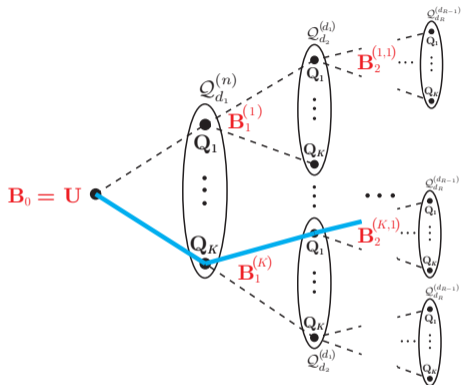
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- Greedy, tree-structured and trellis quantization

S. Schwarz, "Non-coherent broadcasting based on Grassmannian superposition transmission," in *IEEE International Symposium on Information Theory (ISIT)*, July 2021, pp. 1–5, under review

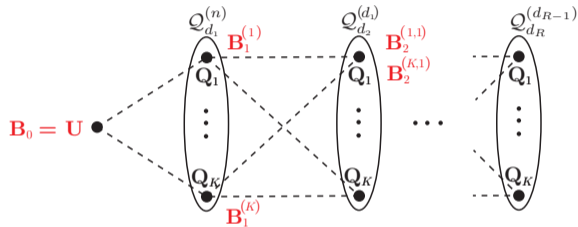
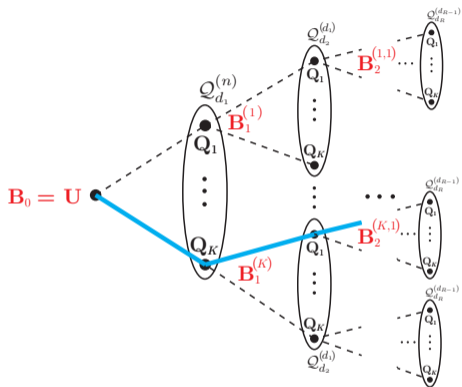
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- Greedy, tree-structured and trellis quantization

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Multi-Stage Manifold Quantization

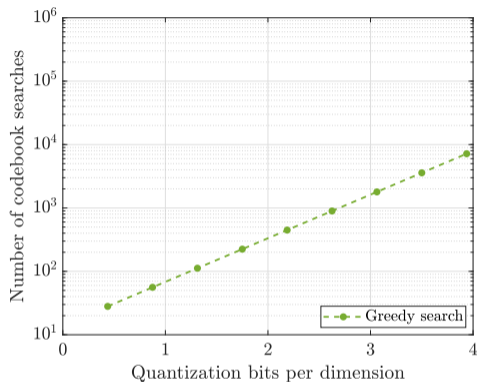
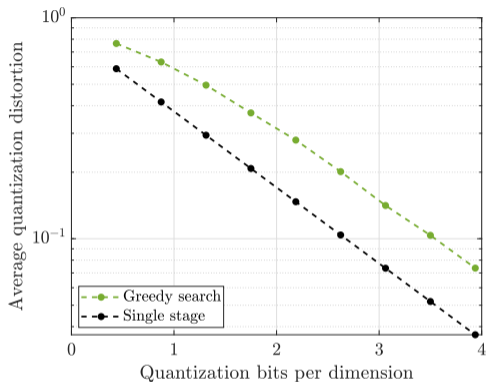
**Performance Evaluation**

Conclusions and Future Work

## Simulation Setup

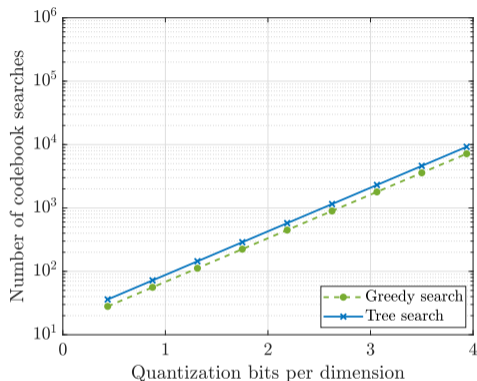
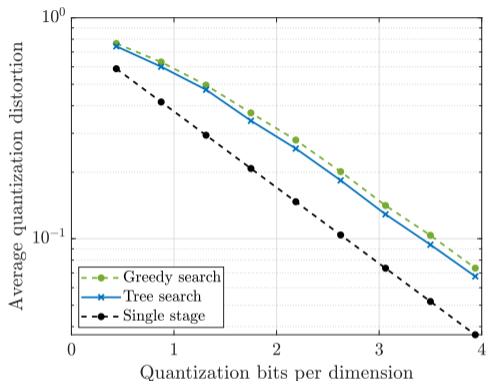
- Quantization of points uniformly distributed on the Grassmannian of dimensions  $n = 16$  and  $m = 2$
- $R = 14$  quantization stages with dimension step-size  $d_{i-1} - d_i = 1$
- Quantization distortion as a function of quantization bits per dimension  $\bar{b} = b/(nm)$  with  $b \in [14, 126]$
- Random vector quantization codebooks
- In the paper: similar results for the Stiefel-manifold

# Greedy, Tree and Trellis Quantization



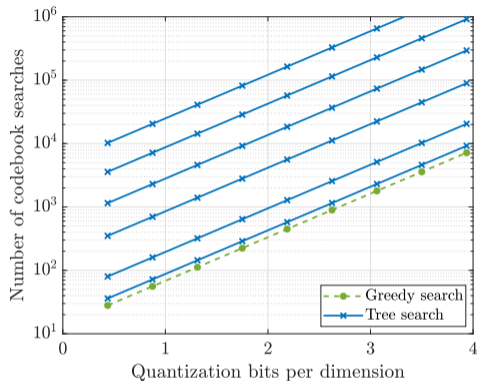
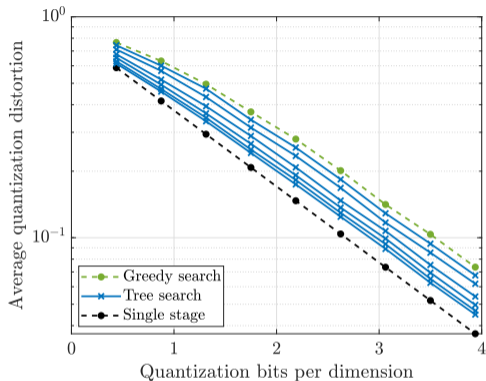
- Complexity of single stage:  $1.6e4$  to  $8.5e37$
- Tree-search of increasing search-depth (by far not up to full tree-search)
- Full and pruned trellis-search

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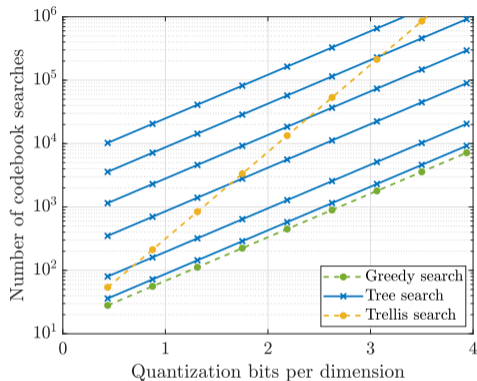
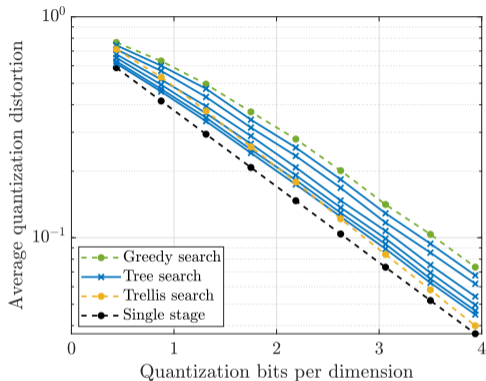
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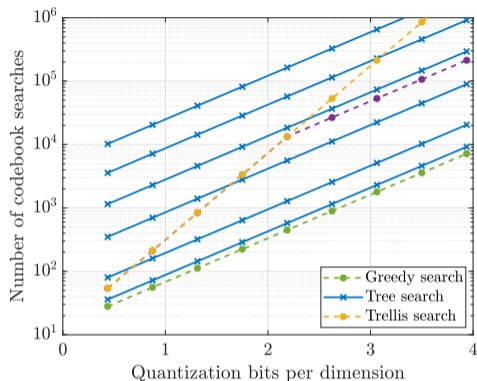
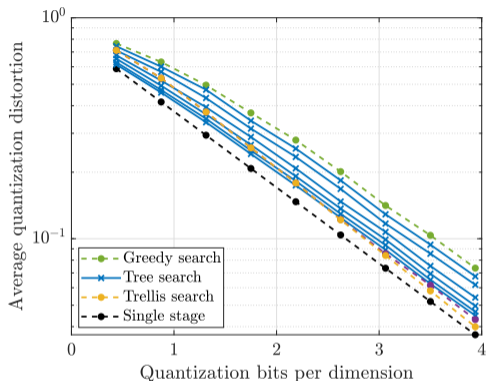
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Multi-Stage Manifold Quantization

Performance Evaluation

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## Conclusions and Future Work

- Recursive multi-stage quantization can support high-resolution with acceptable complexity and a small to moderate performance loss compared to single-stage quantization
- For Grassmannian quantization we provide theoretical rate-distortion results in the paper (assuming random vector quantization)
- Todo: derive similar theoretical results for quantization on the Stiefel manifold



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