# From Robust Adaptive Beamformers to Robust Multi-User MIMO Receivers

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## Introduction

Robust MV beamforming:

- earlier approaches include fixed diagonal loading [Abramovich'81], [Carlson'88]; adaptive diagonal loading [Cox Zeskind Owen'87]; eigenspace-based beamforming [Feldman Griffiths'91], and other techniques.
- More recent approaches are based on *worst-case designs* [Vorobyov Gershman Luo'01], [Lorenz Boyd'01], or can be interpreted by means of such designs [Li Stoica Wang'02].

Our goal: to extend the worst-case MV beamformer designs to multi-user MIMO space-time receivers to make them robust against channel state information (CSI) errors.

(1)

### Traditional and Robust MV Beamforming

Classical (non-robust) MV beamforming:

$$\min_{\mathbf{w}} \mathbf{w}^{H} \mathbf{R} \mathbf{w} \quad \text{s.t.} \quad \mathbf{w}^{H} \mathbf{a}_{s} = 1 \quad \rightarrow \quad \mathbf{w}_{\text{MV}} = (\mathbf{a}_{s}^{H} \mathbf{R}^{-1} \mathbf{a}_{s})^{-1} \mathbf{R}^{-1} \mathbf{a}_{s}$$

Robust worst-case MV beamforming:

$$\min_{\mathbf{w}} \mathbf{w}^{H} \mathbf{R} \mathbf{w} \quad \text{s.t.} \quad |\mathbf{w}^{H} (\mathbf{a}_{s} + \boldsymbol{\delta})| \geq 1 \quad \forall \|\boldsymbol{\delta}\| \leq \varepsilon$$

The latter problem can be transformed to

$$\min_{\mathbf{w}} \mathbf{w}^{H} \mathbf{R} \mathbf{w} \quad \text{s.t.} \quad \mathbf{w}^{H} \mathbf{a}_{s} \ge \varepsilon \|\mathbf{w}\| + 1$$
(2)

Problem (2) is *convex* and can be solved with the complexity  $O(N^3)$  by second-order cone programming (SOCP) algorithms [Vorobyov Gershman Luo'01] or Newton-type algorithms [Lorenz Boyd'01], [Li Stoica Wang'02]



## Array Processing-Type MIMO Model Cont'd

Signal model:

$$\mathbf{Y}_{(T \times M)} = \sum_{p=1}^{P} \mathbf{X}_{p(T \times N)} \mathbf{H}_{(N \times M)} + \mathbf{V}_{(T \times M)}$$

Let the transmitted symbols of the *p*th user be given by the vector  $\mathbf{s}_p \triangleq [s_{p,1} \cdots s_{p,K}]^T$  and let an orthogonal space-time code (OSTBC) [Tarokh Jafarkhani Calderbank'98], [Alamouti'98] be used:

$$\mathbf{X}_p = \mathbf{X}(\mathbf{s}_p), \quad \mathbf{X}^H(\mathbf{s}_p)\mathbf{X}(\mathbf{s}_p) = \|\mathbf{s}_p\|^2 \mathbf{I}_N$$

(3)

## Array Processing-Type MIMO Model Cont'd

The matrix  $\mathbf{X}(\mathbf{s}_p)$  can be written as

$$\mathbf{X}(\mathbf{s}_p) = \sum_{k=1}^{K} \left( \mathbf{C}_k \operatorname{Re}\{s_{p,k}\} + \mathbf{D}_k \operatorname{Im}\{s_{p,k}\} \right)$$

where  $\mathbf{C}_k \triangleq \mathbf{X}(\mathbf{e}_k)$  and  $\mathbf{D}_k \triangleq \mathbf{X}(j\mathbf{e}_k)$ .

Using (3) yields the following *array processing-type* model:

$$\underline{\mathbf{Y}}_{(2MT\times1)} = \sum_{p=1}^{P} \mathbf{A}_{p(2MT\times2K)} \underline{\mathbf{s}}_{p(2K\times1)} + \underline{\mathbf{V}}_{(2MT\times1)}$$
$$\mathbf{A}_{p} = \mathbf{A}(\mathbf{H}_{p}) \triangleq [\underline{\mathbf{C}}_{1}\mathbf{H}_{p} \cdots \underline{\mathbf{C}}_{K}\mathbf{H}_{p} \ \underline{\mathbf{D}}_{1}\mathbf{H}_{p} \cdots \underline{\mathbf{D}}_{K}\mathbf{H}_{p}]$$
$$\triangleq [\mathbf{a}_{p,1} \cdots \mathbf{a}_{p,2K}]$$

where  $\underline{\mathbf{P}} \triangleq [\operatorname{vec} \{\operatorname{Re}(\mathbf{P})\}^T \operatorname{vec} \{\operatorname{Im}(\mathbf{P})\}^T]^T$  for any matrix  $\mathbf{P}$ .

### Linear Receivers

The ML decoder may be prohibitively expensive in the multi-user case. Simpler suboptimal linear receivers can be used:

$$\underline{\hat{\mathbf{s}}_1} = \mathbf{W}^T \underline{\mathbf{Y}}$$

where w.l.g. user # 1 is assumed to be the user-of-interest (UOI) and W is the  $2MT \times 2K$  matrix of receiver coefficients. Example: the matched filter (MF) receiver

$$\underline{\hat{\mathbf{s}}_1} = \frac{1}{\|\mathbf{H}_1\|_F^2} \, \mathbf{A}_1^T \underline{\mathbf{Y}}$$

- optimal in the ML sense in the single user case,
- far away from the optimality in the multi-user case.

### Linear Receivers Cont'd

The MV receiver was used in [Shahbazpanahi et al'05]:

 $\min \mathbf{w}_k^T \mathbf{R} \mathbf{w}_k \quad \text{s.t.} \quad \mathbf{a}_{1,k}^T \mathbf{w}_k = 1 \quad \rightarrow \quad \mathbf{w}_{\mathrm{MV},k} = (\mathbf{a}_{1,k}^H \mathbf{R}^{-1} \mathbf{a}_{1,k})^{-1} \mathbf{R}^{-1} \mathbf{a}_{1,k}$  $\mathbf{W}_k$ where  $\mathbf{R} \triangleq \mathrm{E}\{\underline{\mathbf{Y}}\,\underline{\mathbf{Y}}^T\}$ . Then,  $\mathbf{W}_{\mathrm{MV}} = [\mathbf{w}_{\mathrm{MV},1} \cdots \mathbf{w}_{\mathrm{MV},2K}]$ Self-interference zero-forcing by additional constraints:  $\mathbf{a}_{1l}^T \mathbf{w}_k = 0$  for all  $l \neq k$ Generalized MV (GMV) receiver [Shahbazpanahi et al'05]: min tr{ $\mathbf{W}^T \mathbf{R} \mathbf{W}$ } s.t.  $\mathbf{A}_1^T \mathbf{W} = \mathbf{I}_{2K} \rightarrow \mathbf{W}_{\text{GMV}} = \mathbf{R}^{-1} \mathbf{A}_1 (\mathbf{A}_1^T \mathbf{R}^{-1} \mathbf{A}_1)^{-1}$ How to add robustness against CSI errors at the receive side?





Lemma: For any OSTBC,

$$\|\mathbf{\Delta}\|_F = \|\mathbf{\delta}_k\|$$
 for all  $k = 1, \dots, 2K$ 

where 
$$\boldsymbol{\delta}_k \triangleq \mathbf{a}_k(\mathbf{H}_1) - \mathbf{a}_k(\hat{\mathbf{H}}_1)$$
.

Using this lemma, we can prove that problem (4) takes the following form [Rong Shahbazpanahi Gershman'05]:

$$\min_{\mathbf{w}_k} \mathbf{w}_k^T \mathbf{R} \mathbf{w}_k \quad \text{s.t.} \quad \mathbf{w}_k^T (\mathbf{a}_k(\hat{\mathbf{H}}_1) + \boldsymbol{\delta}_k) \ge 1 \quad \forall \quad \|\boldsymbol{\delta}\| \le \varepsilon$$
(5)

which is *mathematically identical* to the robust MV beamforming problem in (1)! Hence, (5) can be directly solved using the algorithms developed in [Vorobyov Gershman Luo'01], [Lorenz Boyd'01], and [Li Stoica Wang'02].

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## **Robust MV Receivers Cont'd**

- A robust formulation has also been developed for the GMV receiver: more complicated formulation, but still convertible to the SOCP form [Rong Shahbazpanahi Gershman'05].
- As the worst-case designs may be overly conservative, chance programming designs with probabilistic constraints have been proposed in [Rong Vorobyov Gershman'05]:

$$\min_{\mathbf{w}_k} \mathbf{w}_k^T \mathbf{R} \mathbf{w}_k \quad \text{s.t.} \quad \Pr\{\mathbf{w}_k^T(\mathbf{a}_k(\hat{\mathbf{H}}_1) + \boldsymbol{\delta}_k) \ge 1\} > p_o$$

where  $1 - p_o$  can be viewed as the *outage probability*. More flexibility, but still convertible to convex optimization problems. More work on this idea is required.

## **Numerical Examples**

#### Example # 1:

- P = 2 users of N = 2 antennas each; M = 8 receive antennas, Alamouti's code (T = 2, K = 2), QPSK symbols, INR = 20 dB.
- UOI channel matrix is drawn from a zero-mean Gaussian unit-variance distribution.
- UOI channel matrix is known up to CSI errors drawn from a zero-mean Gaussian distribution with variance  $\sigma_e^2 = 0.1$

Example # 2: the same as Example # 1 except that P = 3 users of N = 3 antennas each and Tarokh's 3/4-rate OSTBC (T = 4, K = 3) are assumed.









## Conclusions

- Worst-case robust beamformer designs have been extended to multi-user receivers for orthogonally space-time block coded MIMO systems.
- Further extensions: non-orthogonal high-rate space-time codes, outage probability based formulations, computationally efficient on-line algorithms, combining linear and non-linear receivers.