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# Multibeam Analog Beamformer Design for Monostatic ISAC under Self-Interference

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# ISAC in mmWave Networks

## ■ The ISAC paradigm

- Sharing the hardware platform, signaling, and radio spectrum.
- A core enabler for 6G applications (V2X, UAVs).

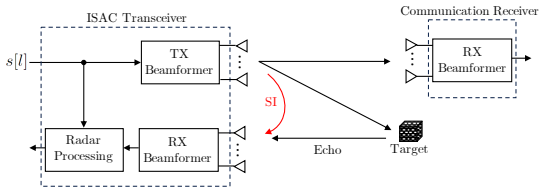
## ■ Suitability of mmWave

- Offers high angular and range resolution.
- Demands directional beamforming to overcome severe path loss.

## ■ The Hardware Reality

- Fully digital architectures are often prohibitive in power and cost.
- Analog beamforming (single RF chain) imposes strict Constant Modulus (CM) constraints.

# The Challenge: Monostatic Operation & Self-Interference



## ■ The Self-Interference (SI) Bottleneck

- The powerful TX signal strongly leaks into the sensing RX array.

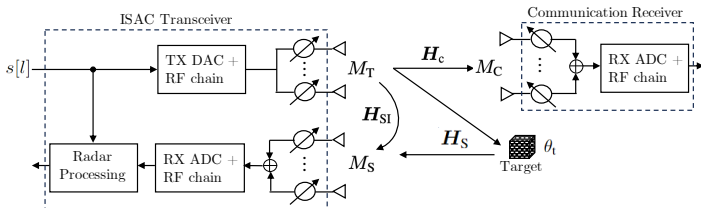
## ■ Hardware Vulnerability & RX Cancellation Limits

- RF cancellation at RX is useful, but cannot protect RF components preceding the canceller.
- Early component saturation ruins any subsequent analog/digital cancellation.

## Our Approach: SI-aware TX Beamforming

Design a TX analog beamformer that maximizes ISAC performance while proactively limiting the SI level at the RX sensing array.

# System Model (I)



- Beamformers:  $\mathbf{f} \in \mathbb{V}^{M_T}$  (TX);  $\mathbf{w}_s \in \mathbb{V}^{M_S}$  (Radar);  $\mathbf{w}_c \in \mathbb{V}^{M_C}$  (RX)

$$\mathbb{V}^m \triangleq \{\mathbf{z} \in \mathbb{C}^m : |z_i| = 1, 1 \leq i \leq m\}$$

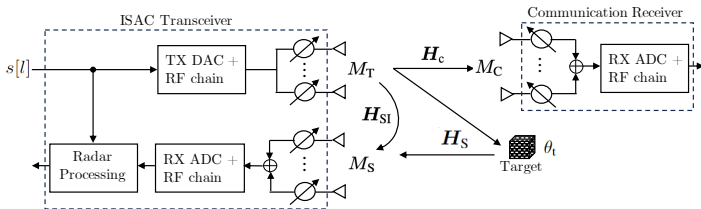
- Each antenna is connected to a PS without individual gain control
- ISAC transceiver sends zero-mean, unit-variance symbols:

$$\mathbf{x}[l] = \mathbf{f}s[l] \in \mathbb{C}^{M_T}$$

- Received baseband communication signal:

$$y[l] = \mathbf{w}_c^H \mathbf{H}_c \mathbf{f} s[l] + \mathbf{w}_c^H \mathbf{n}_c[l] \xrightarrow{\text{for } \mathbf{f} \text{ given}} [\mathbf{w}_c]_i = e^{i\angle[\mathbf{H}_c \mathbf{f}]_i}, \forall i$$

# System Model (and II)



- The radar observation is given by

$$r[l] = \underbrace{\mathbf{w}_s^H \mathbf{H}_S \mathbf{f} s[l]}_{\text{Target's Echo}} + \underbrace{\mathbf{w}_s^H \mathbf{H}_{SI} \mathbf{f} s[l]}_{\text{Post-combining SI}} + \underbrace{\mathbf{w}_s^H \mathbf{n}_s[l]}_{\text{Noise}}$$

- Conventional approach: mitigate Post-Combining SI via  $\{\mathbf{f}, \mathbf{w}_s\}$

Our Approach: Control *Pre-Combining* SI

$$P_{SI} \triangleq \mathbb{E}\{\|\mathbf{H}_{SI} \mathbf{f} s[l]\|^2\} = \mathbf{f}^H \mathbf{H}_{SI}^H \mathbf{H}_{SI} \mathbf{f}$$

# Problem Formulation

## Multibeam SI-aware TX Beamforming

The design of  $\mathbf{f}$  can be addressed as

$$\max_{\mathbf{f}} G_{\text{tx,comm}} \triangleq \mathbf{f}^H \mathbf{H}_c^H \mathbf{H}_c \mathbf{f}$$

subject to:

- Required Beampattern gain:  $G_{\text{tx,sen}}(\theta_t) \triangleq |\mathbf{f}^H \mathbf{a}_T(\theta_t)|^2 \geq \tau^2$
- Controlled Pre-combining SI:  $P_{\text{SI}} \triangleq \mathbf{f}^H \mathbf{H}_{\text{SI}}^H \mathbf{H}_{\text{SI}} \mathbf{f} \leq \eta^2$
- Constant-Modulus entries:  $\mathbf{f} \in \mathbb{V}^{M_T} \implies |f_i| = 1, i \in \{1, \dots, M_T\}$

# Straightforward Approaches

## Direct Projection

- $\tilde{\mathbf{f}}_*$ : relaxed solution obtained by  ~~$\mathbf{f} \in \mathbb{V}^{M_T}$~~   $\rightarrow \|\mathbf{f}\|^2 = M_T$
- The CM solution reads as

$$\mathbf{f}_* = \mathcal{P}_{\mathbb{V}^n} \{\tilde{\mathbf{f}}_*\}$$

where  $\mathcal{P}_{\mathbb{V}^n} \{\mathbf{x}\} : \mathbb{C}^n \rightarrow \mathbb{V}^n$ , so that  $\mathcal{P}_{\mathbb{V}^n} \{\mathbf{x}\} = \left[ \frac{x_1}{|x_1|}, \dots, \frac{x_n}{|x_n|} \right]^T$

## QCQP+SDR

- Square up CM constraints:

$$|f_i| = 1 \rightarrow |f_i|^2 = 1 \iff \mathbf{f}^H \mathbf{e}_i \mathbf{e}_i^H \mathbf{f} = 1, \forall i$$

- Quadratically-Constrained Quadratic Problem (QCQP) with  $M_T + 2$  constraints

→ SDR (generally not tight)

## Proposed Approach

# SI-Aware Analog Multibeam Design

1. Obtain a Semi-Analytic Solution by relaxing CM constraints
2. Modify the Relaxed Solution to handle CM constraints

# Semi-Analytic Relaxed Solution

- Replacing CM constraints by a Total Power Constraint, we have

$$\begin{aligned} \max_{\mathbf{f} \in \mathbb{C}^{M_T}} \quad & G_{\text{tx,comm}} \\ \text{s.t.} \quad & \text{(a) } G_{\text{tx,sen}}(\theta_t) \geq \tau^2, \text{ (b) } P_{\text{SI}} \leq \eta^2, \text{ (c) } \cancel{\mathbf{f} \in \mathbb{V}^{M_T}} \rightarrow \|\mathbf{f}\|^2 = M_T \end{aligned}$$

→ QCQP with 3 constraints

- Structure of the *Relaxed* Optimal Beamformer:

$$\mathbf{f}_\star = \sqrt{M_T} \mathcal{D} \left[ \mathbf{H}_c^H \mathbf{H}_c + \mu_\star \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) - \gamma_\star \mathbf{H}_{\text{SI}}^H \mathbf{H}_{\text{SI}} \right]$$

→  $\mathcal{D}[\mathbf{X}]$ : dominant unit-norm eigenvector of  $\mathbf{X}$

→  $\mu_\star, \gamma_\star$ : optimal Lagrange multipliers for (a) and (b)

$\mu_\star$  and  $\gamma_\star$  do not admit closed-form solution → 3-step procedure

# Relaxed Optimal Beamformer

## 3-step Procedure

### 1. Communication-Optimal Start ( $\mu = \gamma = 0$ )

- Initialize with  $\mathbf{f} = \sqrt{M_T \mathcal{D}} [\mathbf{H}_c^H \mathbf{H}_c]$
- If feasible, stop.

### 2. Single-Constraint Activation ( $\mu = 0$ or $\gamma = 0$ )

- *Communication-Sensing beamformer*  $\mathbf{f}_{C\text{-Sen}}$ : set  $\gamma = 0$ , find  $\mu$
- *Communication-SIC beamformer*  $\mathbf{f}_{C\text{-SIC}}$ : set  $\mu = 0$ , find  $\gamma$
- Bisection Search

$$\mathbf{f}_* = \begin{cases} \mathbf{f}_{C\text{-Sen}} & \text{if only } \mathbf{f}_{C\text{-Sen}} \text{ is feasible} \\ \mathbf{f}_{C\text{-SIC}} & \text{if only } \mathbf{f}_{C\text{-SIC}} \text{ is feasible} \\ \arg \max_{\{\mathbf{f}_{C\text{-Sen}}, \mathbf{f}_{C\text{-SIC}}\}} G_{\text{tx,comm}} & \text{if both are feasible} \end{cases}$$

### 3. Dual-Constraint Optimization ( $\mu \neq 0, \gamma \neq 0$ )

- Optimal values  $\mu_*$  and  $\gamma_*$  must simultaneously satisfy sensing and SI constraints with strict equality

# Restoring Constant-Modulus Constraints

- Starting with the Semi-Analytic Relaxed Solution

$$\mathbf{f}_\star = \sqrt{M_T} \mathcal{D} [\mathbf{H}_c^H \mathbf{H}_c + \mu_\star \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) - \gamma_\star \mathbf{H}_{SI}^H \mathbf{H}_{SI}]$$

- We find the CM-constrained solution

$$\mathbf{f}_\star = \mathcal{P}_{\mathbb{V}^{M_T}} \left\{ \mathcal{D} [\mathbf{H}_c^H \mathbf{H}_c + \mu_\star \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) - \gamma_\star \mathbf{H}_{SI}^H \mathbf{H}_{SI}] \right\}$$

The optimal multipliers are found **iteratively**

## Our Approach

Refine the 3-step procedure to enforce CM constraints **at each iteration**

# Modified 3-step Procedure

## 3-step Procedure

### 1. **Communication-Optimal Start** ( $\mu = \gamma = 0$ )

→ Initialize with

$$\mathbf{f} = \mathcal{P}_{\mathbb{V}^{M_T}} \left\{ \mathcal{D} \left[ \mathbf{H}_c^H \mathbf{H}_c \right] \right\}$$

→ If feasible, stop.

### 2. **Single-Constraint Activation** ( $\mu = 0$ or $\gamma = 0$ )

→ If only one constraint is violated, find the optimal multiplier  $\mu$  or  $\gamma$  via Bisection Search

### 3. **Dual-Constraint Optimization** ( $\mu \neq 0, \gamma \neq 0$ )

→ If both sensing and SI limits are active, refine the solution using a CM-adapted Newton-Raphson method.

# Projected (Constant-Modulus) Bisection

- Let  $\mathbf{X}$  be the objective and  $\mathbf{Y}$  be the constraint
  
- Repeat
  1.  $\omega \leftarrow (\omega_{\min} + \omega_{\max})/2$
  2.  $\mathbf{f} \leftarrow \mathcal{P}_{\mathbb{V}_{M_T}}\{\mathcal{D}[(1 - \omega)\mathbf{X} + \omega\mathbf{Y}]\}$
  3. if  $\mathbf{f}^H \mathbf{Y} \mathbf{f} > \beta$ , then  $\omega_{\max} \leftarrow \omega$
  4. else  $\omega_{\min} \leftarrow \omega$
  
- Until  $|\mathbf{f}^H \mathbf{Y} \mathbf{f} - \beta| \leq \epsilon$

# Projected (Constant-Modulus) Newton-Raphson

■ Repeat:

1.  $\mathbf{u} \leftarrow \mathcal{P}_{\mathbb{V}_{M_T}} \{ \mathcal{D}[(1 - \boldsymbol{\alpha}^T \mathbf{1}) \mathbf{H}_c^H \mathbf{H}_c + \alpha_1 \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) - \alpha_2 \mathbf{H}_{S1}^H \mathbf{H}_{S1}] \}$

2. Set

$$\mathbf{q}(\boldsymbol{\alpha}) \leftarrow \begin{bmatrix} \mathbf{u}^H \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) \mathbf{u} - \tau^2 \\ \mathbf{u}^H \mathbf{H}_{S1}^H \mathbf{H}_{S1} \mathbf{u} - \eta^2 \end{bmatrix}$$

3. Compute the Jacobian matrix

$$\mathbf{J}_q(\boldsymbol{\alpha}) \leftarrow \begin{bmatrix} \frac{\partial}{\partial \alpha[1]} q_1(\boldsymbol{\alpha}) & \frac{\partial}{\partial \alpha[2]} q_1(\boldsymbol{\alpha}) \\ \frac{\partial}{\partial \alpha[1]} q_2(\boldsymbol{\alpha}) & \frac{\partial}{\partial \alpha[2]} q_2(\boldsymbol{\alpha}) \end{bmatrix}$$

4. Update  $\boldsymbol{\alpha} \leftarrow \boldsymbol{\alpha} - \mathbf{J}_q^{-1}(\boldsymbol{\alpha}) \mathbf{q}(\boldsymbol{\alpha})$

5.  $\mathbf{f} \leftarrow \mathcal{P}_{\mathbb{V}_{M_T}} \{ \mathcal{D}[(1 - \boldsymbol{\alpha}^T \mathbf{1}) \mathbf{H}_c^H \mathbf{H}_c + \alpha_1 \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) - \alpha_2 \mathbf{H}_{S1}^H \mathbf{H}_{S1}] \}$

■ Until  $|\mathbf{f}^H \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) \mathbf{f} - \tau^2| \leq \epsilon$  &  $|\mathbf{f}^H \mathbf{H}_{S1}^H \mathbf{H}_{S1} \mathbf{f} - \eta^2| \leq \epsilon$

# Numerical Example (I)

Simulation Setup:

- Communication Channel: narrowband clustered model
- SI Channel: Rice model (Near-Field LoS + Nearby Scatters NLoS)
- Arrays:  $M_T = 32$  (TX),  $M_S = 8$  (Radar),  $M_C = 16$  (Comm. RX)

Sensing Threshold:

$$\tau^2 \in [0, M_T^2]$$

SI Threshold:

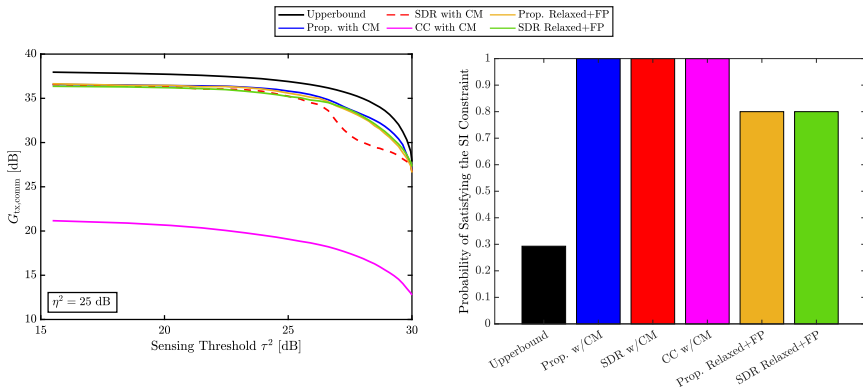
$$\eta^2 = \zeta M_T s_{\max}(\mathbf{H}_{SI}) \leq \zeta M_T \|\mathbf{H}_{SI}\|_F^2 = \zeta M_T^2 M_S \quad \text{with } \zeta < 1$$

Benchmarks:

- Upperbound: unconstrained without SI control (black)
- SDR with CM constraints (red)
- Convex Combination (CC) with CM constraints (magenta)
- Relaxed with Final Projection (orange)
- SDR Relaxed with Final Projection (green)

## Numerical Example (II)

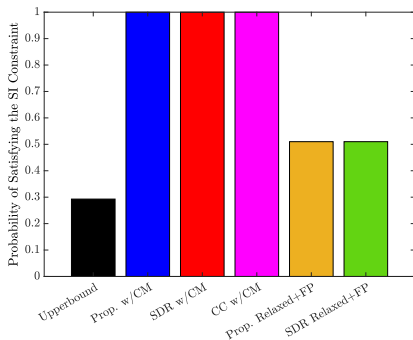
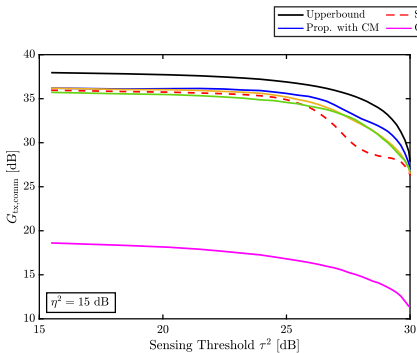
- Moderate SI Threshold (25 dB):  $\eta^2 = \zeta M_T^2 M_S$  with  $\zeta = \frac{1}{25}$



SDR with CM issues: complexity and tightness (60%)

# Numerical Example (and III)

- Strict SI Threshold (15 dB):  $\eta^2 = \zeta M_T^2 M_S$  with  $\zeta = \frac{1}{250}$



SDR with CM issues: complexity and tightness (60%)

# In Summary

- 1 We have addressed the design of multibeam analog beamformers for monostatic ISAC under SI
- 2 With per-antenna gain control (CM constraints relaxed), we have derived a semi-analytic solution in which only the Lagrange multipliers have to be numerically computed
- 3 For CM analog phased arrays, we have enforced CM constraints by projecting the relaxed solution onto the feasible set at each iteration
  - Sensing and SI constraints simultaneously satisfied
  - Performance closed to the upperbound

# Backup: A Detail on Dual-Constraint Optimization

- Find  $\mu, \gamma \in [0, \infty)$  so that both constraints are satisfied with equality

- Procedure:

- Reduce the search space

$$\alpha_1 = \frac{\mu}{1 + \mu + \gamma} \in [0, 1] \quad \text{and} \quad \alpha_2 = \frac{\gamma}{1 + \mu + \gamma} \in [0, 1 - \alpha_1]$$

- Define  $\boldsymbol{\alpha} \triangleq [\alpha_1, \alpha_2]^T$  and

$$\mathbf{u} \triangleq \sqrt{M_T} \mathcal{D}[(1 - \boldsymbol{\alpha}^T \mathbf{1}) \mathbf{H}_c^H \mathbf{H}_c + \alpha_1 \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) - \alpha_2 \mathbf{H}_{SI}^H \mathbf{H}_{SI}]$$

- Find the roots of  $\mathbf{q} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$\mathbf{q}(\boldsymbol{\alpha}) \triangleq \begin{bmatrix} q_1(\boldsymbol{\alpha}) \\ q_2(\boldsymbol{\alpha}) \end{bmatrix} = \begin{bmatrix} \mathbf{u}^H \mathbf{a}_T(\theta_t) \mathbf{a}_T^H(\theta_t) \mathbf{u} - \tau^2 \\ \mathbf{u}^H \mathbf{H}_{SI}^H \mathbf{H}_{SI} \mathbf{u} - \eta^2 \end{bmatrix}$$

via Newton-Raphson