Channel Shaping Using Reconfigurable Intelligent Surfaces: From Diagonal to Beyond

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I. ASSUMPTION

All proposals in this paper based on assumption of asymmetric passive Beyond-Diagonal (BD) Reconfigurable Intelligent Surface (RIS), i.e., symmetry constraint $\Theta_g = \Theta_g^T$ is relaxed. This is feasible when asymmetric passive components (e.g., ring hybrids and branch-line hybrids) [1] are available. This assumption was also made in Hongyu's papers [2], [3]. For quadratic problems, the proposed algorithms may be extended to symmetric BD RIS by replacing singular value decomposition with Takagi factorization [4].

II. POINT-TO-POINT MIMO

A. Channel Power Maximization

Consider a BD RIS with N^{S} elements, which is divided into G groups of equal L elements.

$$\max_{\mathbf{\Theta}} \quad \left\| \mathbf{H}^{\mathrm{D}} + \sum_{g} \mathbf{H}_{g}^{\mathrm{B}} \mathbf{\Theta}_{g} \mathbf{H}_{g}^{\mathrm{F}} \right\|_{\mathrm{F}}^{2}$$
(1a s.t. $\mathbf{\Theta}_{g}^{\mathsf{H}} \mathbf{\Theta}_{g} = \mathbf{I}, \quad \forall g \in \mathcal{G} \triangleq \{1, ..., G\}.$

s.t.
$$\mathbf{\Theta}_g^{\mathsf{H}} \mathbf{\Theta}_g = \mathbf{I}, \quad \forall g \in \mathcal{G} \triangleq \{1, ..., G\}.$$
 (1b)

For symmetric BD-RIS, the problem has been solved in

- Matteo's paper [5]: SISO and equivalent¹;
- Ignacio's paper [6]: SISO and directless MISO/SIMO.

Remark 1. The difficulty of (1) is that the RIS needs to balance the additive (direct-indirect) and multiplicative (forwardbackward) eigenspace alignment. Interestingly, it has the same form as the weighted orthogonal Procrustes problem [7]:

$$\min_{\mathbf{\Theta}} \quad \|\mathbf{C} - \mathbf{A}\mathbf{\Theta}\mathbf{B}\|_{\mathrm{F}}^2 \tag{2a}$$

s.t.
$$\Theta^{\mathsf{H}}\Theta = \mathbf{I}$$
. (2b)

There exists no trivial solution to (2). One lossy transformation, by moving Θ to one side [8], formulates a standard orthogonal Procrustes problem:

$$\min_{\mathbf{\Theta}} \quad \|\mathbf{A}^{\dagger} \mathbf{C} - \mathbf{\Theta} \mathbf{B}\|_{\mathrm{F}}^{2} \tag{3a}$$

s.t.
$$\Theta^{\mathsf{H}}\Theta = \mathbf{I}$$
. (3b)

(3) has a global optimal solution $\Theta^* = UV^H$, where U and V are left and right singular matrix of $A^{\dagger}CB^{H}$ [9]. This low-complexity solution will be compared with the one proposed later.

Inspired by [10], we propose an iterative algorithm to solve (1). The idea is to successively approximate the quadratic

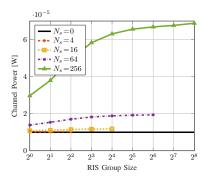


Fig. 1. Average channel power versus RIS elements N^{S} and group size Lfor $(N^{\mathrm{T}}, N^{\mathrm{R}}) = (8,4), (\Lambda^{\mathrm{D}}, \Lambda^{\mathrm{F}}, \Lambda^{\mathrm{B}}) = (65,54,46) \mathrm{dB}.$

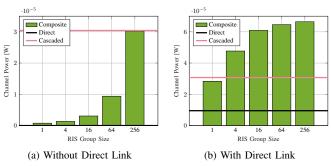


Fig. 2. Average channel power versus RIS group size L for $(N^{\rm T},N^{\rm S},N^{\rm R})=(8,256,4),~(\Lambda^{\rm D},\Lambda^{\rm F},\Lambda^{\rm B})=(65,54,46){\rm dB}.$

objective with a sequence of affine functions and solve the resulting subproblems in closed form.

Proposition 1. Start from any $\Theta^{(0)}$, the sequence

$$\mathbf{\Theta}_{g}^{(r+1)} = \mathbf{U}_{g}^{(r)} \mathbf{V}_{g}^{(r)}, \quad \forall g \in \mathcal{G}$$
 (4)

converges to a stationary point of (1), where $\mathbf{U}_q^{(r)}$ and $\mathbf{V}_q^{(r)}$ are left and right singular matrix of

$$\mathbf{M}_{g}^{(r)} = \mathbf{H}_{g}^{\mathrm{B}\mathsf{H}} \mathbf{H}^{\mathrm{D}} \mathbf{H}_{g}^{\mathrm{F}\mathsf{H}} + \sum_{g' < g} \mathbf{H}_{g'}^{\mathrm{B}\mathsf{H}} \mathbf{H}_{g'}^{\mathrm{B}} \mathbf{\Theta}_{g'}^{(r+1)} \mathbf{H}_{g'}^{\mathrm{F}} \mathbf{H}_{g'}^{\mathrm{F}\mathsf{H}} + \sum_{g' \ge g} \mathbf{H}_{g'}^{\mathrm{B}\mathsf{H}} \mathbf{H}_{g'}^{\mathrm{B}} \mathbf{\Theta}_{g'}^{(r)} \mathbf{H}_{g'}^{\mathrm{F}} \mathbf{H}_{g'}^{\mathrm{F}\mathsf{H}}.$$
(5)

Fig. 1 shows that, apart from adding reflecting elements $N^{\rm S}$, increasing the group size L also improves the channel power. This behavior is more pronounced for a large RIS. For example, the gain of pairwise connection is 2.8% for $N^{\rm S} = 16$

¹Single-stream MIMO with given precoder and combiner.

and 28% for $N^{\rm S}=256$. It implies that the channel shaping capability of BD RIS scales with group size L.

Fig. 2b and 2a compare the average channel power without and with direct link. "Cascaded" means the *power product* of the forward and backward channels. We observe that diagonal RIS wastes substantial cascaded power and struggles to align the direct-indirect eigenspace. When the direct link is absent, only 2.6% of available power is utilized by diagonal RIS while 100% power is recycled by fully-connected RIS. When the direct link is present, the proposed BD RIS design can balance the direct-indirect and forward-backward eigenspace alignment for an optimal channel boost. It is worth noting that, when L is sufficiently large, the composite channel power surpasses the power sum of direct and cascaded channels, thanks to the constructive *amplitude superposition* of direct and cascaded channels. This again emphasizes the advantage of in-group connection of BD RIS.

B. Rate Maximization

The problem is formulated w.r.t. precoder (instead of transmit covariance matrix) for reference:

$$\max_{\mathbf{W}, \mathbf{\Theta}} \quad \operatorname{logdet}\left(\mathbf{I} + \frac{\mathbf{W}^{\mathsf{H}} \mathbf{H}^{\mathsf{H}} \mathbf{H} \mathbf{W}}{\sigma_n^2}\right) \tag{6a}$$

s.t.
$$\|\mathbf{W}\|_{\mathrm{F}}^2 \le P$$
, (6b)

$$\Theta_{a}^{\mathsf{H}}\Theta_{q} = \mathbf{I}, \quad \forall g.$$
 (6c)

(6) is jointly non-convex and solved by Alternating Optimization (AO). For a given Θ , the optimal precoder is given by

$$\mathbf{W}^{\star} = \mathbf{V}\mathbf{S}^{\star 1/2}.\tag{7}$$

where V is right singular matrix of H and S^* is a diagonal matrix of the water-filling power allocation. For a given W, we update Θ by Riemannian Conjugate Gradient (RCG) method along the geodesics [11].

Remark 2. A geodesic refers to the shortest path between two points in a Riemannian manifold. Unitary constraint (6c) translates to a Stiefel manifold where the geodesics have simple expressions described by the exponential map [12].

For optimization problems with block unitary constraint, the adapted RCG method at iteration r for block g is summarized below.

1) Compute the Euclidean gradient

$$\nabla_{\mathbf{\Theta}_{g}}^{\mathbf{E}} f^{(r)} = \frac{\partial f^{(r)}}{\partial \mathbf{\Theta}_{a}^{*}} \tag{8}$$

2) Translate to the Riemannian gradient

$$\nabla^{\mathrm{R}}_{\boldsymbol{\Theta}_g} f^{(r)} \!=\! \nabla^{\mathrm{E}}_{\boldsymbol{\Theta}_g} f^{(r)} \boldsymbol{\Theta_g^{(r)}}^{\mathsf{H}} \!-\! \boldsymbol{\Theta_g^{(r)}} \nabla^{\mathrm{E}}_{\boldsymbol{\Theta}_g} f^{(r)}^{\mathsf{H}} \hspace{0.5cm} (9)$$

3) Determine the weight factor

$$\gamma^{(r)} = \frac{\operatorname{tr}\left(\left(\nabla_{\mathbf{\Theta}_g}^{\mathbf{R}} f^{(r)} - \nabla_{\mathbf{\Theta}_g}^{\mathbf{R}} f^{(r-1)}\right) \nabla_{\mathbf{\Theta}_g}^{\mathbf{R}} f^{(r)\mathsf{H}}\right)}{\operatorname{tr}\left(\nabla_{\mathbf{\Theta}_g}^{\mathbf{R}} f^{(r-1)} \nabla_{\mathbf{\Theta}_g}^{\mathbf{R}} f^{(r-1)\mathsf{H}}\right)}$$
(10)

4) Compute the conjugate direction

$$\mathbf{D}^{(r)} = \nabla_{\mathbf{\Theta}_q}^{\mathbf{R}} f^{(r)} + \gamma^{(r)} \mathbf{D}^{(r-1)}$$
 (11)

5) Determine the Armijo step size

$$\mu^{(r)} = \underset{\mu}{\operatorname{argmax}} f^{(r)} \left(\exp\left(-\mu \mathbf{D}^{(r)}\right) \mathbf{\Theta}_g^{(r)} \right) \tag{12}$$

6) Perform multiplicative (rotational) update

$$\mathbf{\Theta}^{(r+1)} = \exp\left(-\mu^{(r)}\mathbf{D}^{(r)}\right)\mathbf{\Theta}_g^{(r)} \tag{13}$$

The advantage over general manifold optimization [13], [14] is trifold: 1) no retraction is required; 2) no projection is required; 3) no matrix inversion is required.

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