

# D.C programming with rank constrained problem

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## 1 D.C Programming

The D.C programming problems have been widely investigated. It has been proven that D.C programming can only converge to a local optimal stable solution. Now we consider a digital precoder problem:

$$\text{maximize} \quad \sum_{k=1}^K R_k = \sum_{k=1}^K \log_2 \left( \frac{|\mathbf{h}_k \mathbf{f}_k|^2}{\sigma_k^2 + \sum_{i \neq k}^K |\mathbf{h}_k \mathbf{f}_i|^2} \right) \quad (1)$$

$$= \sum_{k=1}^K \left[ \log_2 (|\mathbf{h}_k \mathbf{f}_k|^2) - \log_2 \left( \sigma_k^2 + \sum_{i \neq k}^K |\mathbf{h}_k \mathbf{f}_i|^2 \right) \right]$$

$$\text{subject to} \quad \|\mathbf{V} \mathbf{f}_k\|_F \leq 1;$$

$$R_k \geq \lambda_k, \quad (2)$$

where  $\mathbf{V}$  is the analog precoder and  $\mathbf{h}_k = \mathbf{w}_k^H \mathbf{h}_k \mathbf{V}$  is the effective channel.

Since  $\log_2(\cdot)$  is concave while  $|\mathbf{h}_k \mathbf{f}_k|^2$  is convex, the problem (1) can be transformed as D.C problem by defining:

$$\mathbf{F}_k = \mathbf{f}_k \mathbf{f}_k^H \quad (3)$$

Then the problem (1) can be rewritten as a standard D.C problem:

$$\text{maximize} \quad \sum_{k=1}^K R_k = \sum_{k=1}^K \left[ \log_2 (|\mathbf{h}_k \mathbf{F}_k \mathbf{h}_k^H|) - \log_2 \left( \sigma_k^2 + \sum_{i \neq k}^K |\mathbf{h}_k \mathbf{F}_i \mathbf{h}_k^H| \right) \right] \quad (4)$$

$$\text{subject to} \quad \|\mathbf{V} \mathbf{f}_k\|_F \leq 1;$$

$$R_k \geq \lambda_k;$$

$$\text{Rank}(\mathbf{F}_k) = 1.$$

By ignoring the rank constraint, the digital precoder can be solved alternatively because (1) the degree freedom of variable matrix  $\mathbf{F}_k, k = 1, \dots, K$  will be very large such that it's very hard to be solved; (2) the local optimal solution will be sensitive to initial value as the increment of degree freedom. We firstly give the initial value to  $\mathbf{F}_k^{(0)}, k = 1, \dots, K$ , then the digital precoder is solved iteratively:

$$\text{maximize} \quad f(\mathbf{F}_k) - g(\mathbf{F}_k) : \quad \text{subject to the constraints in (4)} \quad (5)$$

The solution of problem (5) can be calculated by iterative procedure:

$$\text{maximize} \quad f(\mathbf{F}_k) - g(\mathbf{F}_k^{(t)}) - \langle \nabla g(\mathbf{F}_k^{(t)}), \mathbf{F}_k - \mathbf{F}_k^{(t)} \rangle : \quad \text{subject to the constraints in (4)} \quad (6)$$

This convex problem in each iterations can be easily solved by CVX toolbox.

## 2 The rank constrained problem

A general rank constrained problem for SDP variable is formulated as:

$$\begin{aligned} & \text{minimize} && f(\mathbf{X}) \\ & \text{subject to} && \mathbf{X} \succeq 0; \end{aligned} \tag{7}$$

$$\begin{aligned} & \mathbf{x} \in \mathcal{C}, \\ & \text{Rank}(\mathbf{X}) \leq r \end{aligned} \tag{8}$$

where  $\mathcal{C}$  is a convex set and  $f(\mathbf{x})$  is a convex function.

This problem can be solved by an iterative method. At each step  $t$ ,

$$\begin{aligned} & \min_{\mathbf{X}_t, e_t} && f(\mathbf{X}) + w_t e_t \\ & \text{subject to} && \mathbf{x}_t \succeq 0; \\ & && \mathbf{x}_t \in \mathcal{C}, \\ & && e_t \mathbf{I}_{n-r} - \mathbf{U}_{t-1}^T \mathbf{X}_t \mathbf{U}_{t-1} \succeq \mathbf{0}; \\ & && e_t \geq e_{t-1} \end{aligned} \tag{9}$$

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Combining the (9) with (6), the  $F_k$  can be solved by

$$\min_{\mathbf{X}_t, e_t} -f(\mathbf{F}_k) + g(\mathbf{F}_k^{(t)}) + \langle \nabla g(\mathbf{F}_k^{(t)}), \mathbf{F}_k - \mathbf{F}_k^{(t)} \rangle + w_t e_t \tag{10}$$

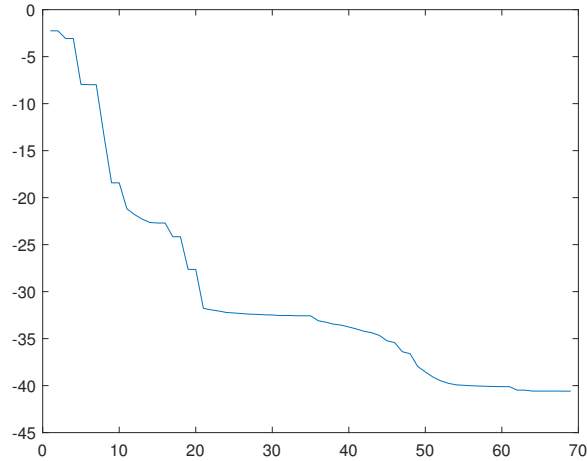


Figure 1: The convergence of proposed algorithm.