



DC programming and DCA for Secure Guarantee with Null Space Beamforming in Two-Way Relay Networks

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01

Introduction

- The problem
- System model

Eavesdropper



Wireless channel

Create secure links relying on **physical characteristic** of wireless channel without relying on the privacy cryptograph

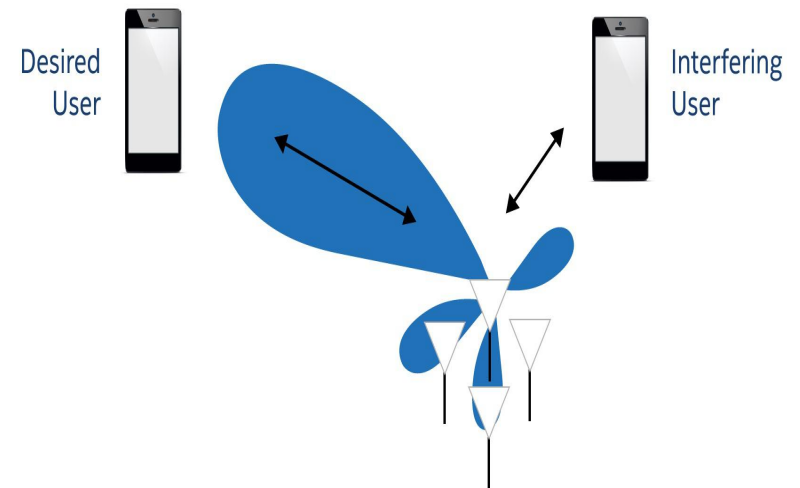
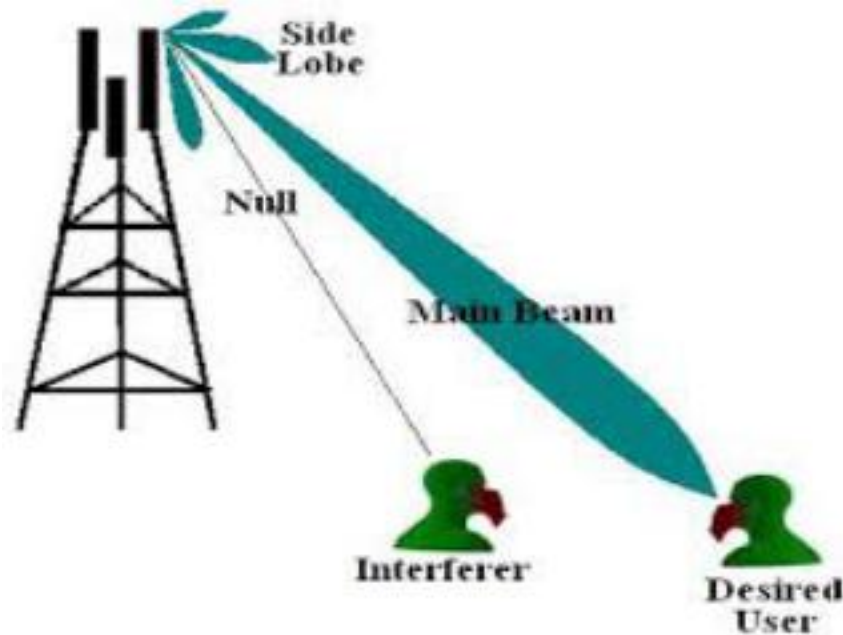


Goals

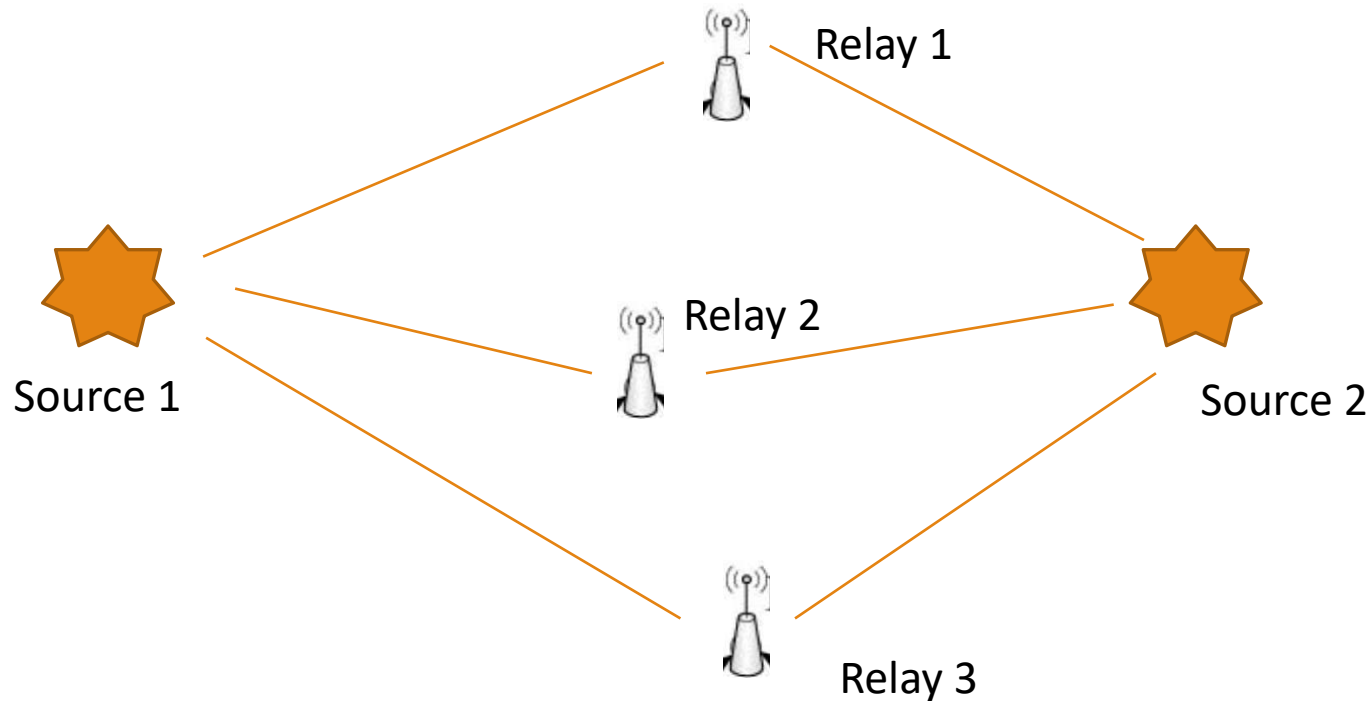
Find **the maximum rate received** at the legitimate receiver, while keeping the **eavesdropper completely ignorant** of the transmitted messages
=> Reduce the probability of interception



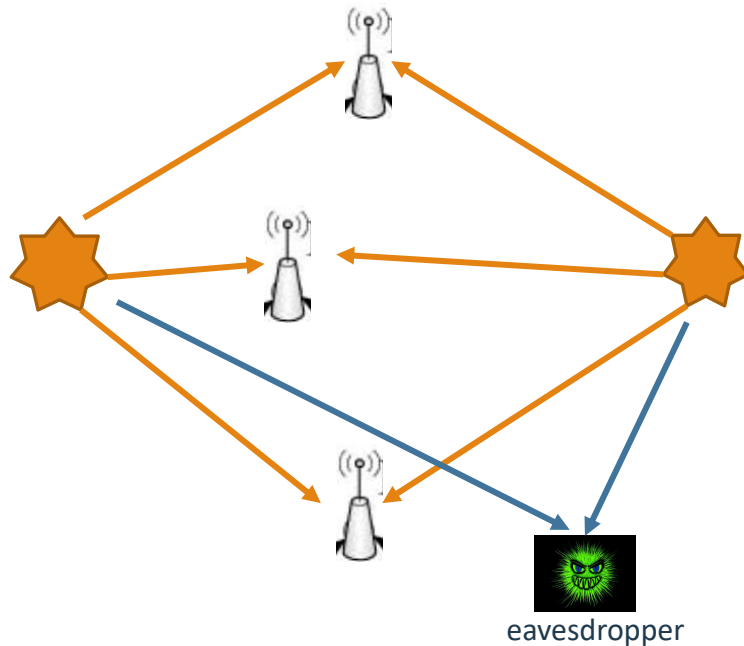
Null-Space Beamforming



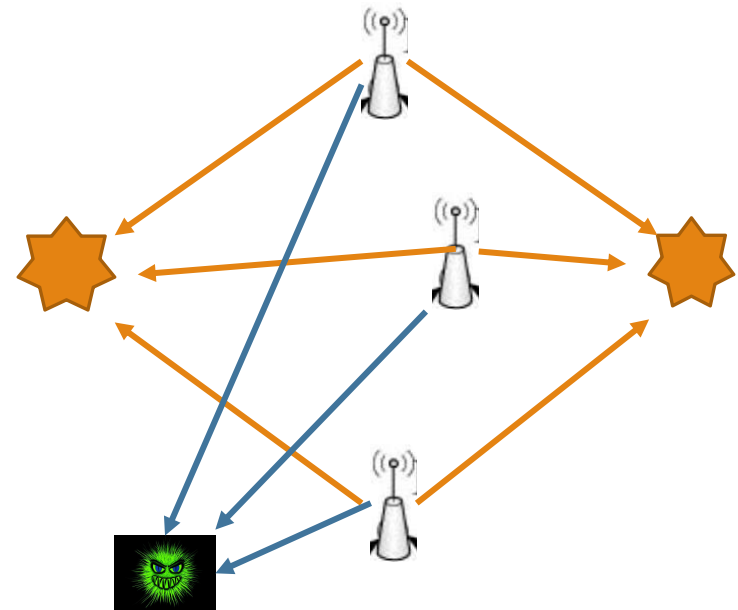
Amplify-and-forward (AF) relaying protocol



Multiple Access Channel (MAC) and Broadcast Channel (BC)



Broadcast Channel



Multiple Access Channel

Model System

$$\max_{P_1, P_2, \mathbf{w}} \quad \frac{1}{2} \left[\log_2 \frac{\left(1 + \frac{P_2 |\mathbf{w}^H \mathbf{F}_1 \mathbf{f}_2|^2}{\sigma_R^2 \|\mathbf{w}^H \mathbf{F}_1\|_2^2 + \sigma_1^2} \right) \left(1 + \frac{P_1 |\mathbf{w}^H \mathbf{F}_2 \mathbf{f}_1|^2}{\sigma_R^2 \|\mathbf{w}^H \mathbf{F}_2\|_2^2 + \sigma_2^2} \right)}{1 + \frac{P_1 |g_1|^2 + P_2 |g_2|^2}{\sigma_R^2}} \right]^+$$

$$\text{s.t} \quad \begin{aligned} P_1 + P_2 + P_1 \mathbf{w}^H \mathbf{D}_1 \mathbf{w} + P_2 \mathbf{w}^H \mathbf{D}_2 \mathbf{w} + \sigma^2 \mathbf{w}^H \mathbf{w} &\leq P_T \\ \mathbf{w}^H \mathbf{Z} &= 0 \end{aligned}$$

²Y. Yang, C. Sun, H. Zhao et al., "Algorithms for secrecy guarantee with null space beamforming in two-way relay networks", *IEEE Trans. Signal Process.*, vol. 62, no. 8, pp. 2111-2126, 2014.



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Optimization of beamforming vector algorithm

Generalized Rayleigh quotient

- For a given pair (\mathbf{A}, \mathbf{B}) of matrices, and a given non-zero vector \mathbf{x} , the generalized Rayleigh quotient is defined as:

$$R(\mathbf{A}, \mathbf{B}, \mathbf{x}) = \frac{\mathbf{x}^T \mathbf{A} \mathbf{x}}{\mathbf{x}^T \mathbf{B} \mathbf{x}}$$

- For any vector \mathbf{x} , one has $R(\mathbf{A}, \mathbf{B}, \mathbf{x}) \in [\lambda_{\min}, \lambda_{\max}]$ where $\lambda_{\min}, \lambda_{\max}$ are the smallest and largest eigenvalue of $\mathbf{B}^{-1}\mathbf{A}$, if $\mathbf{B}^{-1}\mathbf{A}$ is Hermitian matrix.

Generalized Rayleigh quotient

- Step 1 : Transform the model of problem into generalized Rayleigh quotient form.

$$\max_{\mathbf{x} \in \mathbb{C}^{2(N-2) \times 1}} f(\mathbf{x}) = \frac{\mathbf{x}^H (\hat{\mathbf{P}}_1^T \otimes \hat{\mathbf{P}}_2) \mathbf{x}}{\mathbf{x}^H (\hat{\mathbf{Q}}_1^T \otimes \hat{\mathbf{Q}}_2) \mathbf{x}}$$

$$\text{s.t} \quad \mathbf{x}^H \mathbf{x} = 1,$$

$$\text{vec}^{-1}(\mathbf{x}) \succcurlyeq \mathbf{0}, \text{rank}(\text{vec}^{-1}(\mathbf{x})) = 1.$$

Generalized Rayleigh quotient

- Step 1 : Transform the model of problem into generalized Rayleigh quotient form.
- Step 2 : Check if $\mathbf{x}_{\text{opt}} = V_{\max}((\hat{\mathbf{Q}}_1^T \otimes \hat{\mathbf{Q}}_2)^{-1}(\hat{\mathbf{P}}_1^T \otimes \hat{\mathbf{P}}_2))$ is vectorization of Hermitian matrix, then we can find the solution immediately.
- Step 3: If not, using randomization to find approximate solution.

Detail of the transformation in the paper'

$V_{\max}(\mathbf{X})$ is an unit-norm eigenvector of the largest eigenvalue of matrix \mathbf{X}

03

Proposed algorithm

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- Difference of Convex Program
 - DC decomposition for Optimization of Relay Beamforming Vector

Difference of Convex Program

Difference of Convex Functions

In DC approach, the original non-convex function is decomposed to the difference of **two convex functions**.

A standard DC program is of the form as the following:

$$\alpha = \inf \{ f(x) := g(x) - h(x) : x \in S \}$$

A general DC program usually has the form as the following:

$$\alpha = \inf \{ f(x) := g(x) - h(x) : g_i(x) - h_i(x) \leq 0, i = 1, \dots, m \}$$

With g, h being lower semi-continuous proper convex functions on \mathbb{R}^n , S is a convex set.

Difference of Convex Program

Difference of Convex Functions Algorithm

Firstly we decompose the objective function f to a difference of **two convex functions**: $g - h$

Starting from a **primal point** x^0 in \mathbb{R}^n .

With standard DCA, we repeat those steps, for $k = 0, 1, 2, \dots$:

- Compute $y^k \in \partial h(x^k)$

- Compute x^{k+1}

$$x^{k+1} \in \operatorname{argmin}\{g(x) - [h(x^k) + \langle y^k, x - x^k \rangle]\}$$

- Until either $\frac{\|x^{k-1} - x^k\|}{\|x^{k-1}\| + 1} < \epsilon$ or $\frac{|f(x^{k-1}) - f(x^k)|}{|f(x^{k-1})| + 1} < \epsilon$

DC decomposition for Optimization of Relay Beamforming Vector (OBV)

Modify Model System

$$\begin{aligned}
 \min_{P_1, P_2, \mathbf{w}} \quad & F(P_1, P_2, \mathbf{w}) = -\frac{1}{2 \ln(2)} \left[\ln \frac{\frac{\sigma_1^2 + \mathbf{w}^H \mathbf{A}_1 \mathbf{w}}{\sigma_1^2 + \mathbf{w}^H \mathbf{B}_1 \mathbf{w}} \frac{\sigma_1^2 + \mathbf{w}^H \mathbf{A}_1 \mathbf{w}}{\sigma_1^2 + \mathbf{w}^H \mathbf{B}_1 \mathbf{w}}}{1 + \frac{P_1 |g_1|^2 + P_2 |g_2|^2}{\sigma_{E,1}^2}} \right] \\
 \text{s.t} \quad & P_1 + P_2 + P_1 \mathbf{w}^H \mathbf{D}_1 \mathbf{w} + P_2 \mathbf{w}^H \mathbf{D}_2 \mathbf{w} + \sigma^2 \mathbf{w}^H \mathbf{w} \leq P_T \\
 & \mathbf{w}^H \mathbf{Z} = 0
 \end{aligned}$$

DC decomposition for Optimization of Relay Beamforming Vector (OBV)

Real Form

$$\mathbf{x} = [\text{Re}(\mathbf{w})^T \quad \text{Im}(\mathbf{w})^T]^T$$

$$\bar{\mathbf{A}}_1 = \begin{bmatrix} \text{Re}(\mathbf{A}_1) & \text{Im}(\mathbf{A}_1) \\ -\text{Im}(\mathbf{A}_1) & \text{Re}(\mathbf{A}_1) \end{bmatrix},$$

$$\bar{\mathbf{B}}_1 = \begin{bmatrix} \text{Re}(\mathbf{B}_1) & \text{Im}(\mathbf{B}_1) \\ -\text{Im}(\mathbf{B}_1) & \text{Re}(\mathbf{B}_1) \end{bmatrix},$$

$$\bar{\mathbf{Z}} = \begin{bmatrix} \text{Re}(\mathbf{Z}^H) & \text{Im}(\mathbf{Z}^H) \\ -\text{Im}(\mathbf{Z}^H) & \text{Re}(\mathbf{Z}^H) \end{bmatrix},$$

$$\bar{\mathbf{D}}_1 = \begin{bmatrix} \text{Re}(\mathbf{D}_1) & \text{Im}(\mathbf{D}_1) \\ -\text{Im}(\mathbf{D}_1) & \text{Re}(\mathbf{D}_1) \end{bmatrix},$$

$$\bar{\mathbf{A}}_2 = \begin{bmatrix} \text{Re}(\mathbf{A}_2) & \text{Im}(\mathbf{A}_2) \\ -\text{Im}(\mathbf{A}_2) & \text{Re}(\mathbf{A}_2) \end{bmatrix},$$

$$\bar{\mathbf{B}}_2 = \begin{bmatrix} \text{Re}(\mathbf{B}_2) & \text{Im}(\mathbf{B}_2) \\ -\text{Im}(\mathbf{B}_2) & \text{Re}(\mathbf{B}_2) \end{bmatrix},$$

$$\bar{\mathbf{D}}_2 = \begin{bmatrix} \text{Re}(\mathbf{D}_2) & \text{Im}(\mathbf{D}_2) \\ -\text{Im}(\mathbf{D}_2) & \text{Re}(\mathbf{D}_2) \end{bmatrix}$$

DC decomposition for Optimization of Relay Beamforming Vector (OBV)

DC decomposition

$$F(\mathbf{x}) = \frac{1}{2 \ln(2)} [G_2(\mathbf{x}) - H_2(\mathbf{x})]$$

$$G_2(\mathbf{x}) = \frac{1}{2} \rho \|\mathbf{x}\|^2 + \ln(C), \text{ where } C = \frac{P_1 |g_1|^2 + P_2 |g_2|^2}{\sigma_{E,1}^2}$$

$$H_2(\mathbf{x}) = \frac{1}{2} \rho \|\mathbf{x}\|^2 + \ln(\sigma_1^2 + \mathbf{x}^T \bar{\mathbf{A}}_1 \mathbf{x}) + \ln(\sigma_2^2 + \mathbf{x}^T \bar{\mathbf{A}}_2 \mathbf{x}) - \ln(\sigma_1^2 + \mathbf{x}^T \bar{\mathbf{B}}_1 \mathbf{x}) - \ln(\sigma_2^2 + \mathbf{x}^T \bar{\mathbf{B}}_2 \mathbf{x})$$

Where ρ is chosen as the maximum eigenvalue of matrix $\left(\frac{\bar{\mathbf{A}}_1}{2\sigma_1^2} + \frac{\bar{\mathbf{A}}_2}{2\sigma_2^2} + \frac{2\bar{\mathbf{B}}_1}{\sigma_1^2} + \frac{2\bar{\mathbf{B}}_2}{\sigma_2^2} \right)$

DC decomposition for Optimization of Relay Beamforming Vector (OBV)

DC decomposition

Note that **the linear approximation** of $H_2(\mathbf{x})$ at the point \mathbf{x}^k is given by:

$$\bar{H}_2(\mathbf{x}; \mathbf{x}^k) = H_2(\mathbf{x}^k) + \langle \mathbf{y}^k; \mathbf{x} - \mathbf{x}^k \rangle$$

Where

$$\begin{aligned} \mathbf{y}^k = \nabla H_2(\mathbf{x}^k) = & \rho \mathbf{x}^k + \frac{2\bar{\mathbf{A}}_1 \mathbf{x}^k}{\sigma_1^2 + (\mathbf{x}^k)^T \bar{\mathbf{A}}_1 \mathbf{x}^k} + \frac{2\bar{\mathbf{A}}_2 \mathbf{x}^k}{\sigma_2^2 + (\mathbf{x}^k)^T \bar{\mathbf{A}}_2 \mathbf{x}^k} \\ & - \frac{2\bar{\mathbf{B}}_1 \mathbf{x}^k}{\sigma_1^2 + (\mathbf{x}^k)^T \bar{\mathbf{B}}_1 \mathbf{x}^k} - \frac{2\bar{\mathbf{B}}_2 \mathbf{x}^k}{\sigma_2^2 + (\mathbf{x}^k)^T \bar{\mathbf{B}}_2 \mathbf{x}^k} \end{aligned}$$

DC decomposition for Optimization of Relay Beamforming Vector

DCA Scheme

1. Initialization: start with any $\mathbf{x}^0 \in \mathbb{R}^{2N}$, $k = 0$

2. Repeat:

Compute \mathbf{x}^{k+1} by solving the following problem

$$\begin{aligned} \min \quad & G_2(\mathbf{x}) - \bar{\mathbf{H}}_2(\mathbf{x}; \mathbf{x}^k) \\ \text{s.t} \quad & P_1 + P_2 + P_1 \mathbf{x}^T \bar{\mathbf{D}}_1 \mathbf{x} + P_2 \mathbf{x}^T \bar{\mathbf{D}}_2 \mathbf{x} + \sigma_R^2 \mathbf{x}^T \mathbf{x} \leq P_T, \\ & \bar{\mathbf{Z}} \mathbf{x} = 0 \end{aligned}$$

3. Until stopping condition is satisfied, i.e

$$\frac{\|\mathbf{x}^{k+1} - \mathbf{x}^k\|}{\|\mathbf{x}^k\| + 1} < \epsilon \text{ or } \frac{|F(\mathbf{x}^{k+1}) - F(\mathbf{x}^k)|}{|F(\mathbf{x}^k)| + 1} < \epsilon$$



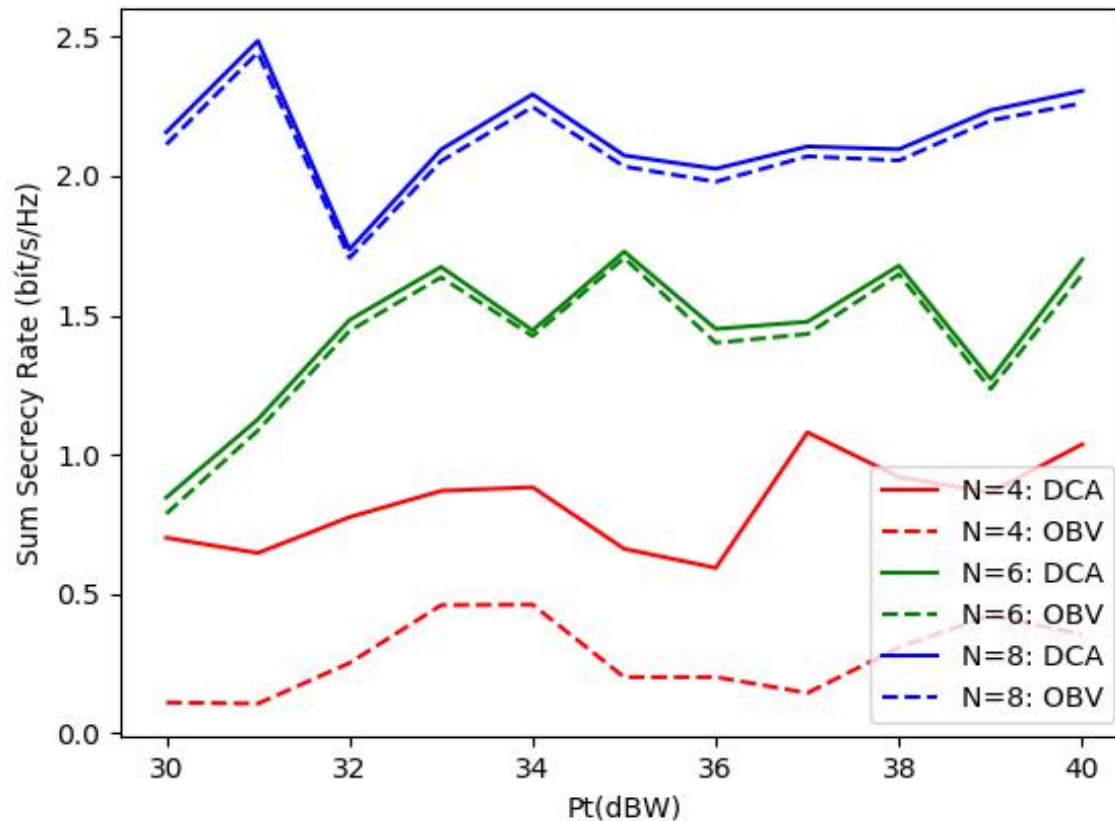
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Experimental results

The analysis part of the comparison **only** focused on the **objective value** of each algorithm.

- Set of power $P_t = [30 : 1 : 40]$ (dBW), the number of relays $N \in \{4, 6, 8\}$, $P_1 = P_2 = \frac{P_T}{4}$ (dBW), $\sigma_1^2 = \sigma_2^2 = 1$ dB.
- Each pair of P_t and N we use 10 sets of randomized data, then getting the average.
- Run on an Intel i7 7500U 2.70 GHz of 8 GB.
- Solver CVX is used for DC algorithm.

3 Experimental results



- There are **a lot of** test sets that OBV gives solution that is **ZERO** while DC algorithm can give **much better results**



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Conclusion and Future Works

Conclusion

- We proposed an algorithm **base on DCA** in order to solve a problem which is finding maximal secrecy sum rate of a two-way relay network system and compare it with an existed algorithm.
- Optimization of Beamforming Vector (OBV) can find quite good solutions in a short amount of time.
- However, DC algorithm even do the job better in a reasonable amount of time.
- While DC algorithm **guarantees convergence**, OBV **cannot** due to the fact that it has a randomization step.

Future Works

- The DC algorithm can be used to improve the objective value of this model in order to get better results for further research.



**Thank you
for
your attention**

Q&A