# Amplify-and-Forward Relay Beamforming for Secrecy with Cooperative Jamming and Imperfect CSI 

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#### Abstract

In this paper, we compute worst case secrecy rates in amplify-and-forward (AF) relay beamforming with cooperative jamming (CJ) in the presence of imperfect channel state information (CSI). A source-destination pair aided by $M$ relays is considered. Number of eavesdroppers $J$ can be more than the number of relays. Out of the $M$ relays, $k_{1}$ relays $\left(1 \leq k_{1} \leq M\right)$ act as data relays and the remaining $k_{2}=M-k_{1}$ relays act as jamming relays. Data relays aid the communication by relaying data in AF mode, and jamming relays cooperate by transmitting jamming signals (artificial noise). The jamming signals are created such that they degrade the eavesdroppers' channels but do not significantly affect the intended receiver's channel, thereby improving secrecy rate. Imperfection in the CSI is modeled using a norm-bounded error model. We solve for the optimum $\left(k_{1}, k_{2}\right)$ and the weights of data relays and jamming relays that maximize the secrecy rate subject to a total relay power constraint. We relax the rank-one constraint on the complex semi-definite data relays and jamming relays weight matrices and reformulate the optimization problem into a form that can be solved using convex semi-definite programming. Numerical results on the secrecy rate that illustrate the effect of cooperative jamming, imperfect CSI, and number of eavesdroppers are presented.


keywords: Cooperative relay beamforming, physical layer security, secrecy rate, amplify-and-forward, cooperative jamming, imperfect CSI, multiple eavesdroppers, semi-definite programming.

## I. Introduction

Wireless transmissions are prone to evesdropping due to their broadcast nature. Providing security through physical layer mechanisms where the intended receiver gets the information reliably while the eavesdroppers get no information is an active area of recent research [1]. Secrecy capacity results for fading channels have been widely reported [2], [3]. Also, secure wireless communications via cooperation is witnessing growing research interest [4]. In particular, cooperation based on amplify-and-forward (AF) and decode-and-forward (DF) relaying protocols for secure communication has been investigated in the literature, assuming perfect and imperfect channel state information (CSI) [5], [6], [7].

Secrecy rate can be improved by adding artificially generated noise (jamming signal) to the information bearing signal such that it degrades the channel towards the eavesdropper but does not degrade the intended receiver's channel [8]. This can be achieved by designing the jamming signal to be orthogonal to the information signal when it reaches the intended receiver, assuming perfect knowledge of CSI. When the sender node has more than one transmit antenna, the additional antennas can be used to transmit jamming signals. Alternately, if 'helper nodes' are available, they can be used to transmit the jamming signals. The idea of helper nodes transmitting jamming signals to improve secrecy rate - referred
to as cooperative jamming (CJ) - has been attracting increased research attention [4], [9]- [12]. Our new contribution in this paper is the evaluation of secrecy rates in an amplify-andforward (AF) relay beamforming scenario in the presence of cooperative jamming and imperfect CSI. Our work is different from the above works on CJ as follows.

In the secrecy rate computation in [4], the following two scenarios are considered. In one scenario, all the $M$ relays are used for relaying data in either AF or DF mode, and there is no cooperative jamming. In the second scenario, all the relays are used for cooperative jamming, and there is no data relaying. These are two extreme cases of the use of relays, which are not necessarily optimal. A more general formulation would be to allow $k_{1}$ out of $M$ relays $\left(1 \leq k_{1} \leq M\right)$ to act as data relays in AF or DF mode and the remaining $k_{2}=M-k_{1}$ relays to act as cooperative jamming relays (see Fig. 1), and solve for the optimum $\left(k_{1}, k_{2}\right)$ that maximizes the secrecy rate. In this paper, we consider this general formulation. In particular, we consider AF protocol for data relaying. We also consider that the knowledge of the CSI is imperfect, and the imperfection is modeled using a norm-bounded CSI error model.

In the above setting, our goal is to solve for the optimum relay beamforming weights (weights of both data relays and jamming relays) that maximize the worst case secrecy rate subject to a total relay power constraint and CSI error constraints. The solution approach adopted is to relax the rankone constraint on the complex semi-definite weight matrices of the data relays and jamming relays and reformulate the optimization problem into a form that can be solved using convex semi-definite programming.

## II. System Model

Consider the cooperative relay beamforming system model shown in Fig. 1, which consists of a source node $S, M$ relay nodes $\left\{R_{1}, R_{2}, \cdots, R_{M}\right\}$, an intended destination node $D$, and $J$ eavesdropper nodes $\left\{E_{1}, E_{2}, \cdots, E_{J}\right\}$, where $J$ can be greater than $M$ (i.e., more number of eavesdroppers than the number of relays). In addition to the links from relays to destination node and relays to eavesdropper nodes, we assume direct links from source to destination node and source to eavesdropper nodes. The complex fading channel gains between source to relays are denoted by $\left\{\gamma_{1}^{*}, \gamma_{2}^{*}, \cdots, \gamma_{M}^{*}\right\}$. Likewise, the channel gains between relays to destination and relays to the $j$ th eavesdropper are denoted by $\left\{\alpha_{1}^{*}, \alpha_{2}^{*}, \cdots, \alpha_{M}^{*}\right\}$ and $\left\{\beta_{1 j}^{*}, \beta_{2 j}^{*}, \cdots, \beta_{M j}^{*}\right\}$, respectively, where $j=1, \cdots, J$. The channel gains on the direct links from source to destination and source to $j$ th eavesdropper are


Fig. 1. Relay beamforming with data relays and cooperative jamming relays.
denoted by $\alpha_{0}^{*}$ and $\beta_{0 j}^{*}$, respectively. The channel gains are assumed to be i.i.d. complex Gaussian with zero mean and variances $\sigma_{\gamma_{i}^{*}}^{2}, \sigma_{\alpha_{0}^{*}}^{2}, \sigma_{\alpha_{i}^{*}}^{2}, \sigma_{\beta_{0 j}^{*}}^{2}$, and $\sigma_{\beta_{i j}^{*}}^{2}$. Let $P_{0}$ denote the total transmit power budget in the system (i.e., source power plus relays power).

## A. Relay Beamforming Using AF with Cooperative Jamming

The source transmits data in the first hop of transmission. Let $x$ be the source symbol transmitted from the source with $\mathbb{E}\left\{|x|^{2}\right\}=1$. In the second hop of transmission, let $k_{1}$ out of $M$ relays, $1 \leq k_{1} \leq M$, are selected to act as the data relays to aid communication from $S$ to $D$ and remaining $k_{2}=M-k_{1}$ relays are selected to act as jamming relays for the transmission of jamming signals. Let $P_{s}$ denote the power transmitted by the source in the first hop of transmission. Let $\left\{\phi_{1}, \phi_{2}, \cdots, \phi_{k_{1}}\right\}$ denote the complex weights applied on the transmitted signals from the $k_{1}$ data relays and let $\left\{\psi_{1}, \psi_{2}, \cdots, \psi_{k_{2}}\right\}$ denote the complex weights applied on the transmitted jamming signals from the $k_{2}$ jamming relays in the second hop of transmission. Let $\boldsymbol{y}_{R}^{k_{1}}, y_{D_{1}}$ and $y_{E_{1 j}}$ denote the received signals at the $k_{1}$ data relays, destination $D$ and $j$ th eavesdropper $E_{j}$, respectively, in the first hop of transmission. In the second hop of transmission, the received signals at the destination and $j$ th eavesdropper are denoted by $y_{D_{2}}$ and $y_{E_{2 j}}$, respectively. We have

$$
\begin{array}{r}
\boldsymbol{y}_{R}^{k_{1}}=\sqrt{P_{s}} \gamma^{k_{1} *} x+\boldsymbol{\eta}_{R}^{k_{1}}, \\
y_{D_{1}}=\sqrt{P_{s}} \alpha_{0}^{*} x+\eta_{D_{1}}, \\
y_{E_{1 j}}=\sqrt{P_{s}} \beta_{0 j}^{*} x+\eta_{E_{1 j}}, \quad j=1, \cdots, J, \\
y_{D_{2}}=\boldsymbol{y}_{R}^{k_{1} T} \operatorname{diag}\left(\boldsymbol{\phi}^{k_{1}}\right) \boldsymbol{\alpha}^{k_{1} *}+\boldsymbol{\alpha}^{k_{2} \dagger} \operatorname{diag}\left(\boldsymbol{\psi}^{k_{2}}\right) \boldsymbol{z}^{k_{2}}+\eta_{D_{2}} \\
=\sqrt{P_{s}} \boldsymbol{\gamma}^{k_{1} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}} x+\boldsymbol{\eta}_{R}^{k_{1} T} \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}}+ \\
\boldsymbol{\alpha}^{k_{2} \dagger} \operatorname{diag}\left(\boldsymbol{\psi}^{k_{2}}\right) \boldsymbol{z}^{k_{2}}+\eta_{D_{2}}, \\
y_{E_{2 j}}=\boldsymbol{y}_{R}^{k_{1} T} \operatorname{diag}\left(\boldsymbol{\phi}^{k_{1}}\right) \boldsymbol{\beta}_{j}^{k_{1} *}+\boldsymbol{\beta}_{j}^{k_{2} \dagger} \operatorname{diag}\left(\boldsymbol{\psi}^{k_{2}}\right) \boldsymbol{z}^{k_{2}}+\eta_{E_{2 j}} \\
=\sqrt{P_{s}} \boldsymbol{\gamma}^{k_{1} \dagger} \operatorname{diag} \operatorname{dig}_{j}^{\dagger}\left(\boldsymbol{\beta}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}} x+\boldsymbol{\eta}_{R}^{k_{1} T} \operatorname{diag}^{\dagger}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}}+ \\
\boldsymbol{\beta}_{j}^{k_{2} \dagger} \operatorname{diag}\left(\boldsymbol{\psi}^{k_{2}}\right) \boldsymbol{z}^{k_{2}}+\eta_{E_{2 j}}, \forall j=1, \cdots, J, \tag{5}
\end{array}
$$

where $\boldsymbol{\gamma}^{k_{1} *}=\left[\gamma_{1}^{*}, \cdots, \gamma_{k_{1}}^{*}\right]^{T}, \boldsymbol{\eta}_{R}^{k_{1}}=\left[\eta_{R_{1}}, \cdots, \eta_{R_{k_{1}}}\right]^{T}, \boldsymbol{\phi}^{k_{1}}=$ $\left[\phi_{1}, \cdots, \phi_{k_{1}}\right]^{T}, \boldsymbol{\psi}^{k_{2}}=\left[\psi_{1}, \cdots, \psi_{k_{2}}\right]^{T}, \boldsymbol{\alpha}^{k_{1} *}=\left[\alpha_{1}^{*}, \cdots, \alpha_{k_{1}}^{*}\right]^{T}$, $\boldsymbol{\alpha}^{k_{2} *}=\left[\alpha_{\left(k_{1}+1\right)}^{*}, \cdots, \alpha_{M}^{*}\right]^{T}, \boldsymbol{\beta}_{j}^{k_{1} *}=\left[\beta_{1 j}^{*}, \cdots, \beta_{k_{1}}^{*}\right]^{T}, \boldsymbol{\beta}_{j}^{k_{2} *}=$ $\left[\beta_{\left(k_{1}+1\right) j}^{*}, \cdots, \beta_{M j}^{*}\right]^{T}, j=1, \cdots, J, \boldsymbol{z}^{k_{2}}=\left[z_{1}, \cdots, z_{k_{2}}\right]^{T}$, and $[.]^{T},(.)^{*},[.]^{\dagger}$ denote transpose, conjugate, conjugate transpose operations, respectively. The noise components, $\eta$ 's and $z$ 's are assumed to be i.i.d. $\mathcal{C N}\left(0, N_{0}\right)$ and $\mathcal{C N}(0,1)$, respectively. Also, $\eta$ 's and $z$ 's are assumed to be independent.

1) Secrecy Rate with Perfect CSI: Let $R_{D}^{k_{1} k_{2}}$ and $R_{E_{j}}^{k_{1} k_{2}}$ denote the information rates at the destination $D$ and $j$ th eavesdropper $E_{j}$, respectively. Using (2) and (4), the expression for the information rate at the destination $D$ is

$$
\begin{equation*}
R_{D}^{k_{1} k_{2}}=\frac{1}{2} \log _{2}\left(1+\frac{P_{s} \alpha_{0}^{*} \alpha_{0}}{N_{0}}+\frac{P_{s} t_{11}}{t_{12}}\right) \tag{6}
\end{equation*}
$$

where

$$
\begin{array}{r}
t_{11}=\boldsymbol{\phi}^{k_{1} \dagger} \operatorname{diag}\left(\boldsymbol{\alpha}^{k_{1}}\right) \boldsymbol{\gamma}^{k_{1}} \boldsymbol{\gamma}^{k_{1} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}} \\
t_{12}=N_{0}+\boldsymbol{\phi}^{k_{1} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{1}}\right) \operatorname{diag}\left(\boldsymbol{\alpha}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}}+ \\
\boldsymbol{\psi}^{k_{2} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{2}}\right) \operatorname{diag}\left(\boldsymbol{\alpha}^{k_{2}}\right) \boldsymbol{\psi}^{k_{2}} .
\end{array}
$$

Similarly, using (3) and (5), the expression for the information rate at the $j$ th eavesdropper $E_{j}$ is

$$
\begin{equation*}
R_{E_{j}}^{k_{1} k_{2}}=\frac{1}{2} \log _{2}\left(1+\frac{P_{s} \beta_{0 j}^{*} \beta_{0 j}}{N_{0}}+\frac{P_{s} t_{21}}{t_{22}}\right) \tag{7}
\end{equation*}
$$

where

$$
\begin{gathered}
t_{21}=\boldsymbol{\phi}^{k_{1} \dagger} \operatorname{diag}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right) \boldsymbol{\gamma}^{k_{1}} \boldsymbol{\gamma}^{k_{1} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}} \\
t_{22}=N_{0}+N_{0} \boldsymbol{\phi}^{k_{1} \dagger} \operatorname{diag}{ }^{\dagger}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right) \operatorname{diag}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}}+ \\
\boldsymbol{\psi}^{k_{2} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\beta}_{j}^{k_{2}}\right) \operatorname{diag}\left(\boldsymbol{\beta}_{j}^{k_{2}}\right) \boldsymbol{\psi}^{k_{2}}, \forall j=1, \cdots, J .
\end{gathered}
$$

The total power transmitted by all $M$ relays is $P_{s} \phi^{k_{1} \dagger} \operatorname{diag}\left(\boldsymbol{\gamma}^{k_{1}}\right) \operatorname{diag}^{\dagger}\left(\boldsymbol{\gamma}^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}}+N_{0} \boldsymbol{\phi}^{k_{1} \dagger} \boldsymbol{\phi}^{k_{1}}+\boldsymbol{\psi}^{k_{2} \dagger} \boldsymbol{\psi}^{k_{2}}$. The achievable secrecy rate $R_{s}^{k_{1} k_{2}}$ for $\left(k_{1}, k_{2}\right)$ relay combination is [4]:

$$
\begin{array}{r}
R_{s}^{k_{1} k_{2}}=\max _{\boldsymbol{\phi}^{k_{1}}, \boldsymbol{\psi}^{k_{2}}} \min _{j: 1, \cdots, J}\left(R_{D}^{k_{1} k_{2}}-R_{E_{j}}^{k_{1} k_{2}}\right) \\
=\max _{\boldsymbol{\phi}^{k_{1}}, \boldsymbol{\psi}^{k_{2}}} \min _{j: 1, \cdots, J} \frac{1}{2} \log _{2} t_{3} \\
P_{s} \boldsymbol{\phi}^{k_{1} \dagger} \operatorname{diag}\left(\gamma^{k_{1}}\right) \operatorname{diag}^{\dagger}\left(\gamma^{k_{1}}\right) \boldsymbol{\phi}^{k_{1}}+N_{0} \boldsymbol{\phi}^{k_{1} \dagger} \boldsymbol{\phi}^{k_{1}} \\
+\boldsymbol{\psi}^{k_{2} \dagger} \boldsymbol{\psi}^{k_{2}} \leq P_{0}-P_{s}  \tag{9}\\
\text { where } \quad t_{3}=\frac{1+\frac{P_{s} \alpha_{0}^{*} \alpha_{0}}{N_{0}}+\frac{P_{s} t_{11}}{t_{12}}}{1+\frac{P_{s} \beta_{0 j}^{*} \beta_{0 j}}{N_{0}}+\frac{P_{s} t_{21}}{t_{22}}}
\end{array}
$$

Defining $\boldsymbol{\Phi}^{k_{1}} \triangleq \boldsymbol{\phi}^{k_{1}} \boldsymbol{\phi}^{k_{1} \dagger}$ and $\boldsymbol{\Psi}^{k_{2}} \triangleq \boldsymbol{\psi}^{k_{2}} \boldsymbol{\psi}^{k_{2} \dagger}$, the above secrecy rate expression can be written in the following equivalent optimization form:

$$
\begin{equation*}
R_{s}^{k_{1} k_{2}}=\max _{\boldsymbol{\Phi}^{k_{1}}, \boldsymbol{\Psi}^{k_{2}}} \min _{j: 1, \cdots, J} \frac{1}{2} \log _{2} t_{4} \tag{10}
\end{equation*}
$$

s.t. $\quad \boldsymbol{\Phi}^{k_{1}} \succeq 0, \operatorname{rank}\left(\boldsymbol{\Phi}^{k_{1}}\right)=1, \boldsymbol{\Psi}^{k_{2}} \succeq 0, \operatorname{rank}\left(\boldsymbol{\Psi}^{k_{2}}\right)=1$,

$$
\begin{align*}
& \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}\left(\gamma^{k_{1}}\right) \operatorname{diag}^{\dagger}\left(\boldsymbol{\gamma}^{k_{1}}\right)+N_{0} \boldsymbol{I}\right)\right) \\
& +\operatorname{tr}\left(\Psi^{k_{2}}\right) \leq P_{0}-P_{s},  \tag{11}\\
& \text { where } t_{4}=\frac{\left(1+\frac{P_{s} \alpha_{0}^{*} \alpha_{0}}{N_{0}}+\frac{P_{s} t_{411}}{t_{12}}\right)}{\left(1+\frac{P_{s} \beta_{0 j}^{0} \beta_{0 j}}{N_{0}}+\frac{P_{s} t_{421}}{t_{422}}\right)}, \\
& t_{411}=\operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}} \operatorname{diag}\left(\boldsymbol{\alpha}^{k_{1}}\right) \boldsymbol{\gamma}^{k_{1}} \boldsymbol{\gamma}^{k_{1} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{1}}\right)\right), \\
& t_{412}=N_{0}+N_{0} \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}} \operatorname{diag}\left(\boldsymbol{\alpha}^{k_{1}}\right) \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{1}}\right)\right) \quad+ \\
& \operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}} \operatorname{diag}\left(\boldsymbol{\alpha}^{k_{2}}\right) \operatorname{diag}^{\dagger}\left(\boldsymbol{\alpha}^{k_{2}}\right)\right), \\
& t_{421}=\operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}} \operatorname{diag}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right) \boldsymbol{\gamma}^{k_{1}} \boldsymbol{\gamma}^{k_{1} \dagger} \operatorname{diag}^{\dagger}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right)\right), \\
& t_{422}=N_{0}+N_{0} \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}} \operatorname{diag}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right) \operatorname{diag}^{\dagger}\left(\boldsymbol{\beta}_{j}^{k_{1}}\right)\right) \quad+ \\
& \operatorname{tr}\left(\Psi^{k_{2}} \operatorname{diag}\left(\boldsymbol{\beta}_{j}^{k_{2}}\right) \operatorname{diag}^{\dagger}\left(\boldsymbol{\beta}_{j}^{k_{2}}\right)\right) .
\end{align*}
$$

Relaxing the rank constraint on $\boldsymbol{\Phi}^{k_{1}}, \boldsymbol{\Psi}^{k_{2}}$ and dropping the logarithm [5]- [7], [10], [13], the optimization problem to compute the above secrecy rate expression can be written in the following equivalent optimization form:

$$
\begin{array}{r}
\max _{\boldsymbol{\Phi}^{k_{1}}, \boldsymbol{\Psi}^{k_{2}}} \min _{j: 1, \cdots, J} t_{4} \\
\text { s.t. } \quad \boldsymbol{\Phi}^{k_{1}} \succeq 0, \quad \boldsymbol{\Psi}^{k_{2}} \succeq 0, \\
\operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}\left(\gamma^{k_{1}}\right) \operatorname{diag}\left(\gamma^{k_{1}}\right)+N_{0} \boldsymbol{I}\right)\right)+ \\
\operatorname{tr}\left(\mathbf{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s} . \tag{13}
\end{array}
$$

## B. Imperfect CSI Model

We consider imperfect CSI, modeled as $\gamma_{i}^{*}=\widehat{\gamma}_{i}^{*}+e_{\gamma_{i}^{*}}^{*}, i=$ $1, \cdots, k_{1}, \alpha_{0}^{*}=\widehat{\alpha}_{0}^{*}+e_{\alpha_{0}^{*}}^{*}, \beta_{0 j}^{*}=\widehat{\beta}_{0 j}^{*}+e_{\beta_{0 j}^{*}}^{*}, j=1, \cdots, J$, $\boldsymbol{\alpha}^{k_{1} *}=\widehat{\boldsymbol{\alpha}}^{k_{1} *}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1} *}}, \boldsymbol{\alpha}^{k_{2} *}=\widehat{\boldsymbol{\alpha}}^{k_{2} *}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2} *}}, \boldsymbol{\beta}_{j}^{k_{1} *}=\widehat{\boldsymbol{\beta}}_{j}^{k_{1} *}+$ $\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}}^{*}, j=1, \cdots, J, \boldsymbol{\beta}_{j}^{k_{2} *}=\widehat{\boldsymbol{\beta}}_{j}^{k_{2} *}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2}}}^{*}, j=1, \cdots, J$, where $\gamma_{i}^{*}$ 's, $\alpha_{0}^{*}, \beta_{0 j}^{*}$ 's, $\boldsymbol{\alpha}^{k_{1} *}, \boldsymbol{\alpha}^{k_{2} *}, \boldsymbol{\beta}_{j}^{k_{1} *}$ 's, $\boldsymbol{\beta}_{j}^{k_{2} *}$ 's are the true CSI, $\widehat{\gamma}_{i}^{*}$ 's, $\widehat{\alpha}_{0}^{*}, \widehat{\beta}_{0 j}^{*}$ 's, $\widehat{\boldsymbol{\alpha}}^{k_{1} *}, \widehat{\boldsymbol{\alpha}}^{k_{2} *} \widehat{\boldsymbol{\beta}}_{j}^{k_{1} *}$ 's, $\widehat{\boldsymbol{\beta}}_{j}^{k_{2} *}$,s are the corresponding imperfect CSI, and $e_{\gamma_{i}^{*}}^{*}$ 's, $e_{\alpha_{0}^{*}}^{*}, e_{\beta_{0 j}^{*}}^{*}$ 's, $e_{\alpha^{k_{1}}}^{*}$, $\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2}}}^{*}, \boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}}^{*}$ 's, $\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2} *}}^{*}$ 's are the additive errors in the CSI. We consider a norm-bounded CSI error model, where it is assumed that

$$
\begin{array}{r}
\left|e_{\gamma_{i}^{*}}\right| \leq \epsilon_{\gamma_{i}^{*}},\left|e_{\alpha_{0}^{*}}\right| \leq \epsilon_{\alpha_{0}^{*}},\left|e_{\beta_{0 j}^{*}}\right| \leq \epsilon_{\beta_{0 j}^{*}},\left\|\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1} *}}\right\| \leq \epsilon_{\boldsymbol{\alpha}^{k_{1} *}}, \\
\quad\left\|\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2} *}}\right\| \leq \epsilon_{\boldsymbol{\alpha}^{k_{2} *}},\left\|\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}}\right\| \leq \epsilon_{\boldsymbol{\beta}_{j}^{k_{1} *}},\left\|\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2} *}}\right\| \leq \epsilon_{\boldsymbol{\beta}_{j}^{k_{2} *}} .
\end{array}
$$

In addition to the above assumptions, we assume the knowledge of the combined channel gain vectors $(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}=$ $\left[\gamma_{1}^{*} \alpha_{1}^{*}, \cdots, \gamma_{k_{1}}^{*} \alpha_{k_{1}}^{*}\right]^{T}$ and $(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}=\left[\gamma_{1}^{*} \beta_{1 j}^{*}, \cdots, \gamma_{k_{1}}^{*} \beta_{k_{1} j}^{*}\right]^{T}$, $j=1, \cdots, J$. In support of this assumption, we note that estimates of the product channel gains $(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}$ and $(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}$,s can be obtained using the techniques proposed in [14], [15]. The imperfections in the knowledge of these channel vectors are modeled as $(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}=(\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1} *}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}}^{*}$ and $(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}=$ $(\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1} *}+\boldsymbol{e}_{(\gamma \boldsymbol{\beta})_{j}^{k_{1} *}}^{*}$, where $(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}$ and $(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}$ are the perfect CSI, $(\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1} *}$ and $(\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1} *}$ are the corresponding imperfect CSI, and $\boldsymbol{e}_{(\gamma \boldsymbol{\alpha})^{k_{1} *}}^{*}$ and $\boldsymbol{e}_{(\gamma \boldsymbol{\beta})_{j}^{k_{1} *}}^{*}$ are the additive errors in the CSI. A norm-bounded error model is assumed, i.e., $\left\|\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1}{ }^{*}}}\right\| \leq \epsilon_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}}$ and $\left\|\boldsymbol{e}_{(\gamma \boldsymbol{\gamma})_{j}^{k_{1} *}}\right\| \leq \epsilon_{(\gamma \boldsymbol{\beta})_{j}^{k_{1}}}$, $j=1, \cdots, J$.

## III. SECRECY Rate for AF Relay Beamforming with CJ AND IMPERFECT CSI

The optimization problem (12) for computing the worst case secrecy rate of AF relay beamforming can be written in the following form:
where $\widehat{\boldsymbol{\gamma}}^{k_{1} *}=\left[\widehat{\gamma}_{1}^{*}, \cdots, \widehat{\gamma}_{k_{1}}^{*}\right]^{T}, e_{\gamma^{k_{1} *}}^{*}=\left[e_{\gamma_{1}^{*}}^{*}, \cdots, e_{\gamma_{k_{1}}^{*}}^{*}\right]^{T}$,

$$
t_{5}=\frac{1+\frac{P_{s}\left(\widehat{\alpha}_{0}+e_{\alpha_{0}^{*}}\right)^{*}\left(\widehat{\alpha}_{0}+e_{\alpha_{0}^{*}}\right)}{N_{0}}+\frac{P_{s} t_{511}}{t_{512}}}{1+\frac{P_{s}\left(\widehat{\beta}_{0 j}+e_{\beta_{0 j}^{*}}\right)^{*}\left(\widehat{\beta}_{0 j}+e_{\beta_{0 j}^{*}}\right)}{N_{0}}+\frac{P_{s} t_{521}}{t_{522}}},
$$

$$
t_{511}=\left((\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1}}}\right)^{\dagger} \boldsymbol{\Phi}^{k_{1}}\left((\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1}}}\right)
$$

$$
t_{512}=N_{0}+N_{0}\left(\widehat{\boldsymbol{\alpha}}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1} *}}\right)^{\dagger} \boldsymbol{A}^{k_{1}}\left(\widehat{\boldsymbol{\alpha}}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1} *}}\right)+
$$

$$
\left(\widehat{\boldsymbol{\alpha}}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2} *}}\right)^{\dagger} \boldsymbol{B}^{k_{2}}\left(\widehat{\boldsymbol{\alpha}}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2}}}\right)
$$

$$
\left.t_{521}=\left((\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}}\right)^{\dagger} \boldsymbol{\Phi}^{k_{1}}(\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}}\right)
$$

$$
t_{522}=N_{0}+N_{0}\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}}\right)^{\dagger} \boldsymbol{A}^{k_{1}}\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1}}}\right)+
$$

$$
\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2}}}\right)^{\dagger} \boldsymbol{B}^{k_{2}}\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2} *}}\right)
$$

$\boldsymbol{A}^{k_{1}} \triangleq \operatorname{diag}\left(\operatorname{diag}\left(\boldsymbol{\Phi}^{k_{1}}\right)\right)$ and $\boldsymbol{B}^{k_{2}} \triangleq \operatorname{diag}\left(\operatorname{diag}\left(\boldsymbol{\Psi}^{k_{2}}\right)\right)$.

$$
\begin{equation*}
\text { Let } \quad a=\min _{e_{\alpha_{0}^{*}}}\left(1+\frac{P_{s}\left(\widehat{\alpha}_{0}+e_{\alpha_{0}^{*}}\right)^{*}\left(\widehat{\alpha}_{0}+e_{\alpha_{0}^{*}}\right)}{N_{0}}\right) \tag{16}
\end{equation*}
$$

Using the method of Lagrangian, the dual of the above problem is the following SDP problem [16]:
$a=\max _{\eta, \nu} \eta$
s.t. $\nu \geq 0,\left[\begin{array}{cl} & \\ \left(\frac{P_{s}}{N_{0}}+\nu\right) & \frac{P_{s}}{N_{0}} \widehat{\alpha}_{0} \\ \widehat{\alpha}_{0}^{*} \frac{P_{s}}{N_{0}} & \widehat{\alpha}_{0}^{*} \frac{P_{s}}{N_{0}} \widehat{\alpha}_{0}+1-\eta-\nu \epsilon_{\alpha_{0}^{*}}^{2}\end{array}\right] \succeq 0$.
Let $b_{j}=\max _{e_{\beta_{0 j}^{*}}^{*}}\left(\begin{array}{l}\left.1+\frac{P_{s}\left(\widehat{\beta}_{0 j}+e_{\beta_{0 j}^{*}}\right)^{*}\left(\widehat{\beta}_{0 j}+e_{\beta_{0 j}^{*}}\right)}{N_{0}}\right), \\ \forall j: 1, \cdots, J \quad \text { s.t. } \quad\left|e_{\beta_{0 j}^{*}}\right| \leq \epsilon_{\beta_{0 j}^{*}} .\end{array}\right.$
The above problem can be written in the following SDP form:

$$
\begin{gather*}
b_{j}=-\max _{\eta, \nu} \eta, \quad \forall j: 1, \cdots, J \quad \text { s.t. } \nu \geq 0,  \tag{22}\\
{\left[\begin{array}{cc}
\left(-\frac{P_{s}}{N_{0}}+\nu\right) & -\frac{P_{s}}{N_{0}} \widehat{\beta}_{0 j} \\
-\widehat{\beta}_{0 j}^{*} \frac{P_{s}}{N_{0}} & -\widehat{\beta}_{0 j}^{*} \frac{P_{s}}{N_{0}} \widehat{\beta}_{0 j}-1-\eta-\nu \epsilon_{\beta_{0 j}^{*}}^{2}
\end{array}\right] \succeq 0 .} \tag{23}
\end{gather*}
$$

Since the objective function of the optimization problem (14) is independent of $e_{\gamma_{i}^{*}}$, we modify the power constraint

$$
\begin{align*}
& \text { s.t. } \forall i: 1, \cdots, k_{1},\left|e_{\gamma_{i}^{*}}\right| \leq \epsilon_{\gamma_{i}^{*}}\left|e_{\alpha_{0}^{*}}\right| \leq \epsilon_{\alpha_{0}^{*}} \text {, } \\
& \left\|\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1}{ }^{*}}}\right\| \leq \epsilon_{\boldsymbol{\alpha}^{k_{1}}},\left\|\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2}}}\right\| \leq \epsilon_{\boldsymbol{\alpha}^{k_{2}}},\left\|\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1}{ }^{*}}}\right\| \leq \epsilon_{(\boldsymbol{\gamma})^{k_{1}}}, \\
& \left|e_{\beta_{0 j}^{*}}\right| \leq \epsilon_{\beta_{0 j}^{*}},\left\|\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}}\right\| \leq \epsilon_{\boldsymbol{\beta}_{j}^{k_{1} *}},\left\|\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2}}}\right\| \leq \epsilon_{\boldsymbol{\beta}_{j}^{k_{2}}}, \\
& \left\|\boldsymbol{e}_{(\boldsymbol{\gamma \beta})_{j}^{k_{1}}}\right\| \leq \epsilon_{(\boldsymbol{\gamma \beta})_{j}^{k_{1}{ }^{*}}}, \boldsymbol{\Phi}^{k_{1}} \succeq 0, \boldsymbol{\Psi}^{k_{2}} \succeq 0, \\
& \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}\left(\widehat{\boldsymbol{\gamma}}^{k_{1}}+e_{\boldsymbol{\gamma}^{k_{1} *}}\right) \operatorname{diag}^{\dagger}\left(\widehat{\boldsymbol{\gamma}}^{k_{1}}+e_{\boldsymbol{\gamma}^{k_{1} *}}\right)+N_{0} \boldsymbol{I}\right)\right) \\
& +\operatorname{tr}\left(\Psi^{k_{2}}\right) \leq P_{0}-P_{s}, \tag{15}
\end{align*}
$$

$$
\begin{array}{r}
\operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}\left(\widehat{\boldsymbol{\gamma}}^{k_{1}}+e_{\gamma^{k_{1} *}}\right) \operatorname{diag}^{\dagger}\left(\widehat{\gamma}^{k_{1}}+e_{\gamma^{k_{1}}}\right)+N_{0} \boldsymbol{I}\right)\right) \\
+\quad \operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s} \\
\left|e_{\gamma_{i}^{*}}\right| \leq \epsilon_{\gamma_{i}^{*}}, \forall i: 1, \cdots, k_{1},
\end{array}
$$

which is a function of $e_{\gamma_{i}^{*}}$, as follows:

$$
\begin{array}{r}
\max _{\gamma_{\gamma_{i}^{*}}} \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}\left(\widehat{\gamma}^{k_{1}}+e_{\gamma^{k_{1}}}\right) \operatorname{diag}^{\dagger}\left(\widehat{\gamma}^{k_{1}}+e_{\gamma^{k_{1}}}\right)+N_{0} \boldsymbol{I}\right)\right) \\
+\quad \operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s} \\
\text { s.t. } \quad\left|e_{\gamma_{i}^{*}}\right| \leq \epsilon_{\gamma_{i}^{*}}, \forall i: 1, \cdots, k_{1}
\end{array}
$$

Rewriting it in the following equivalent form, we have

$$
\begin{equation*}
\operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}(\boldsymbol{v})+N_{0} \boldsymbol{I}\right)\right)+\operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s} \tag{24}
\end{equation*}
$$

where $\boldsymbol{v}=\left[v_{1}, v_{2}, \cdots, v_{k_{1}}\right]^{T}$, and $\forall i: 1, \cdots, k_{1}$,

$$
v_{i}=\max _{e_{\gamma_{i}^{*}}^{*}}\left(\widehat{\gamma}_{i}+e_{\gamma_{i}^{*}}\right)^{*}\left(\widehat{\gamma}_{i}+e_{\gamma_{i}^{*}}\right) \quad \text { s.t. } \quad\left|e_{\gamma_{i}^{*}}\right| \leq \epsilon_{\gamma_{i}^{*}}
$$

which can be written in the following SDP form:

$$
\begin{array}{r}
v_{i}=-\max _{\eta, \nu} \eta, \quad \forall i: 1, \cdots, k_{1} \\
\text { s.t. } \nu \geq 0, \quad\left[\begin{array}{cc}
(-1+\nu) \\
-\widehat{\gamma}_{i}^{*} & -\widehat{\gamma}_{i}^{*} \widehat{\gamma}_{i}-\eta-\nu \epsilon_{\gamma_{i}}^{2}
\end{array}\right] \succeq 0 .
\end{array}
$$

Substituting the values of $a$ and $b_{j}, j: 1, \cdots, J$ and the power constraint (24) in the optimization problem (14), we have

$$
\begin{aligned}
& \text { s.t. } \quad \boldsymbol{\Phi}^{k_{1}} \succeq 0, \boldsymbol{\Psi}^{k_{2}} \succeq 0,
\end{aligned}
$$

$$
\begin{align*}
& \left\|e_{(\gamma \alpha)^{k_{1} *}}\right\| \leq \epsilon_{(\gamma \alpha)^{k_{1}+}},\left\|e_{\beta_{j}^{k_{1}^{*}}}\right\| \leq \epsilon_{\mathcal{\beta}_{j}^{k_{1}},} \\
& \left\|e_{\boldsymbol{\beta}_{j}^{k_{2}+}}\right\| \leq \epsilon_{\boldsymbol{\beta}_{j}^{k_{2}+}},\left\|e_{(\gamma \boldsymbol{\beta})_{j}^{k_{1}^{*}}}\right\| \leq \epsilon_{(\gamma \boldsymbol{\beta})_{j}^{k_{1 *}}}, \\
& \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}(\boldsymbol{v})+N_{0} \boldsymbol{I}\right)\right)+\operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s}, \tag{26}
\end{align*}
$$

We transform the innermost minimization in the optimizatio problem (25) as
with the constraints written in the equivalent LMIs (Linear Matrix Inequalities) using S-procedure [16] as follows:

$$
\begin{aligned}
& \forall \boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1}}} \quad \text { s.t. }\left\|\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1}}}\right\| \leq \epsilon_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}} \Longrightarrow \\
& -\left((\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1} *}}\right)^{\dagger} \boldsymbol{\Phi}^{k_{1}}\left((\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1 *}}}\right)+t_{6} \leq 0 \\
& {\left[\begin{array}{cc}
\boldsymbol{\Phi}^{k_{1}}+\lambda \boldsymbol{I} & \Longleftrightarrow \lambda \geq 0, \quad \boldsymbol{C}_{\lambda} \triangleq \\
(\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1} \dagger} \boldsymbol{\Phi}^{k_{1}} & (\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1} \dagger} \boldsymbol{\Phi}^{k_{1}}(\widehat{\boldsymbol{\gamma} \boldsymbol{\alpha}})^{k_{1}}-t_{6}-\lambda \epsilon_{(\boldsymbol{\gamma} \boldsymbol{\alpha})^{k_{1}}{ }^{k_{1} \dagger}\left(\widehat{\widehat{\boldsymbol{\alpha}}}{ }^{k_{1}}\right.} \quad[\succeq 0,
\end{array}\right.} \\
& \forall \boldsymbol{e}_{\boldsymbol{\alpha}^{k} 1^{*}} \quad \text { s.t. }\left\|\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1}}}\right\| \leq \epsilon_{\boldsymbol{\alpha}^{k} 1^{*}} \Longrightarrow \\
& N_{0}\left(\widehat{\boldsymbol{\alpha}}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1} *}}\right)^{\dagger} \boldsymbol{A}^{k_{1}}\left(\widehat{\boldsymbol{\alpha}}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{1} *}}\right)+N_{0}-t_{7} \leq 0 \\
& {\left[\begin{array}{cc}
-N_{0} \boldsymbol{A}^{k_{1}}+\mu \boldsymbol{I} & \Longleftrightarrow \mu \geq 0, \quad \boldsymbol{C}_{\mu} \triangleq \\
-N_{0} \widehat{\boldsymbol{\alpha}}^{k_{1} \dagger} \boldsymbol{A}^{k_{1}} & -N_{0} \widehat{\boldsymbol{\alpha}}^{k_{1} \dagger} \boldsymbol{A}^{k_{1}} \widehat{\boldsymbol{\alpha}}^{k_{1}}-N_{0} \boldsymbol{A}^{k_{1} \dagger} \widehat{\boldsymbol{\alpha}}^{k_{1}}+t_{7}-\mu \epsilon_{(\boldsymbol{\alpha})^{k_{1} *}}^{2}
\end{array}\right] \succeq 0,} \\
& \forall \boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2}}} \text { s.t. }\left\|\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2}}}\right\| \leq \epsilon_{\boldsymbol{\alpha}^{k_{2}}} \Longrightarrow \\
& \left(\widehat{\boldsymbol{\alpha}}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2} *}}\right)^{\dagger} \boldsymbol{B}^{k_{2}}\left(\widehat{\boldsymbol{\alpha}}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\alpha}^{k_{2}}}\right)-t_{8} \leq 0 \\
& \Longleftrightarrow \quad \xi \geq 0, \quad \boldsymbol{C}_{\xi} \triangleq \\
& {\left[\begin{array}{cc}
-\boldsymbol{B}^{k_{2}}+\xi \boldsymbol{I} & -\boldsymbol{B}^{k_{2} \dagger} \widehat{\boldsymbol{\alpha}}^{k_{2}} \\
-\widehat{\boldsymbol{\alpha}}^{k_{2} \dagger} \boldsymbol{B}^{k_{2}} & -\widehat{\boldsymbol{\alpha}}^{k_{2} \dagger} \boldsymbol{B}^{k_{2}} \widehat{\boldsymbol{\alpha}}^{k_{2}}+t_{8}-\xi \epsilon_{(\boldsymbol{\alpha})^{k_{2} *}}^{2}
\end{array}\right] \succeq 0,} \\
& \forall \boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}} \text { s.t. }\left\|\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}}\right\| \leq \epsilon_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}} \Longrightarrow \\
& \left((\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}}\right)^{\dagger} \boldsymbol{\Phi}^{k_{1}}\left((\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1}}+\boldsymbol{e}_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}}\right)-t_{9 j} \leq 0 \\
& \Longleftrightarrow \quad \lambda_{j} \geq 0, \quad \boldsymbol{C}_{\lambda_{j}} \triangleq \\
& {\left[\begin{array}{cc}
-\boldsymbol{\Phi}^{k_{1}}+\lambda_{j} \boldsymbol{I} & \Longleftrightarrow \boldsymbol{\Phi}_{j}^{k_{1} \dagger}(\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1}} \\
-(\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1} \dagger} \boldsymbol{\Phi}^{k_{1}} & -(\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1} \dagger} \boldsymbol{\Phi}^{k_{1}}(\widehat{\boldsymbol{\gamma} \boldsymbol{\beta}})_{j}^{k_{1}}+t_{9 j}-\lambda_{j} \epsilon_{(\boldsymbol{\gamma} \boldsymbol{\beta})_{j}^{k_{1} *}}^{2}
\end{array}\right] \succeq 0,} \\
& \forall \boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}} \text { s.t. }\left\|\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}}\right\| \leq \epsilon_{\boldsymbol{\beta}_{j}^{k_{1} *}} \Longrightarrow \\
& -N_{0}\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1} *}}\right)^{\dagger} \boldsymbol{A}^{k_{1}}\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{1}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{1 *}}}\right)-N_{0}+t_{10 j} \leq 0 \\
& \Longleftrightarrow \quad \mu_{j} \geq 0, \quad \boldsymbol{C}_{\mu_{j}} \triangleq \\
& {\left[\begin{array}{cc}
N_{0} \boldsymbol{A}^{k_{1}}+\mu_{j} \boldsymbol{I} & N_{0} \boldsymbol{A}^{k_{1} \dagger} \widehat{\boldsymbol{\beta}}_{j}^{k_{1}} \\
N_{0} \widehat{\boldsymbol{\beta}}_{j}^{k_{1} \dagger} \boldsymbol{A}^{k_{1}} & N_{0} \widehat{\boldsymbol{\beta}}_{j}^{k_{1} \dagger} \boldsymbol{A}^{k_{1}} \widehat{\boldsymbol{\beta}}_{j}^{k_{1}}+N_{0}-t_{10 j}-\mu_{j} \epsilon_{\boldsymbol{\beta}_{j}^{k_{1} *}}^{2}
\end{array}\right] \succeq 0,} \\
& \forall \boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2}}} \text { s.t. }\left\|\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2}}}\right\| \leq \epsilon_{\boldsymbol{\beta}_{j}^{k_{2} *}} \Longrightarrow \\
& -\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2} *}}\right)^{\dagger} \boldsymbol{B}^{k_{2}}\left(\widehat{\boldsymbol{\beta}}_{j}^{k_{2}}+\boldsymbol{e}_{\boldsymbol{\beta}_{j}^{k_{2} *}}\right)+t_{11 j} \leq 0 \\
& \begin{array}{rc} 
& \Longleftrightarrow \quad \xi_{j} \geq 0, \quad \boldsymbol{C}_{\xi_{j}} \triangleq \\
{\left[\begin{array}{cc}
\boldsymbol{B}^{k_{2}}+\xi_{j} \boldsymbol{I} & \boldsymbol{B}^{k_{2} \dagger} \widehat{\boldsymbol{\beta}}_{j}^{k_{2}} \\
\widehat{\boldsymbol{\beta}}_{j}^{k_{2} \dagger} \boldsymbol{B}^{k_{2}} & \widehat{\boldsymbol{\beta}}_{j}^{k_{2} \dagger} \boldsymbol{B}^{k_{2}} \widehat{\boldsymbol{\beta}}_{j}^{k_{2}}-t_{11 j}-\xi_{j} \epsilon_{\boldsymbol{\beta}_{j}^{k_{2} *}}^{2}
\end{array}\right] \succeq 0,}
\end{array} \\
& \text { and } \quad t_{6} \geq 0, t_{10 j} \geq 0, t_{11 j} \geq 0 \text {. }
\end{aligned}
$$

Substituting the maximization form (27) and the above LMI constraints in the optimization problem (25) and using the fact that min-max is greater the max-min, we get the following lower bound for the optimization problem (25):

$$
\begin{equation*}
\max _{\boldsymbol{\Phi}^{k_{1}, \Psi^{k_{2}}}} \max _{\substack{t_{6}, t_{7}, t_{8}, t_{9 j}, t_{10}, t_{1 j}, \lambda, \mu, \xi, \lambda_{j}, \mu_{j}, \xi_{j}, \forall j: 1, \ldots, J}} \min _{j: 1, \cdots, J} \frac{a+P_{s} \frac{t_{6}}{t_{7}+t_{8}}}{b_{j}+P_{s} \frac{t_{9 j}}{t_{10 j}+t_{11 j}}} \tag{28}
\end{equation*}
$$

$$
\begin{array}{r}
\text { s.t. } \quad \boldsymbol{\Phi}^{k_{1}} \succeq 0, \boldsymbol{\Psi}^{k_{2}} \succeq 0, t_{6} \geq 0, \\
\operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}(\boldsymbol{v})+N_{0} \boldsymbol{I}\right)\right)+\operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s}, \\
\lambda \geq 0, \mu \geq 0, \xi \geq 0, \boldsymbol{C}_{\lambda} \succeq 0, \boldsymbol{C}_{\mu} \succeq 0, \quad \boldsymbol{C}_{\xi} \succeq 0, \\
\forall j: 1, \cdots, J, \quad t_{10 j} \geq 0, t_{11 j} \geq 0, \\
\lambda_{j} \geq 0, \mu_{j} \geq 0, \quad \xi_{j} \geq 0, \boldsymbol{C}_{\lambda_{j}} \succeq 0, \boldsymbol{C}_{\mu_{j}} \succeq 0, \boldsymbol{C}_{\xi_{j}} \succeq 0 . \tag{29}
\end{array}
$$

Transforming the innermost minimization in (28) to maximization operation, the optimization problem reduces to the following single maximization form:

$$
\begin{align*}
& \text { max }  \tag{30}\\
& \begin{array}{c}
\boldsymbol{\Phi}^{k_{1}}, \Psi^{k_{2}}, t_{6}, t_{7}, t_{8}, t_{9 j}, t_{10 j}, t_{111}, \\
\lambda, \mu, \xi, \lambda_{j}, \mu_{j}, \xi_{j}, r, s, \forall j: 1, \cdots, j
\end{array} \\
& \lambda, \mu, \xi, \lambda_{j}, \mu_{j}, \xi_{j}, r, s, \forall j: 1, \cdots, J \\
& \text { s.t. } \quad \boldsymbol{\Phi}^{k_{1}} \succeq 0, \boldsymbol{\Psi}^{k_{2}} \succeq 0, t_{6} \geq 0, \\
& \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}(\boldsymbol{v})+N_{0} \boldsymbol{I}\right)\right)+\operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s}, \\
& a\left(t_{7}+t_{8}\right)+P_{s} t_{6}-r\left(t_{7}+t_{8}\right) \geq 0, \\
& \lambda \geq 0, \mu \geq 0, \xi \geq 0, \boldsymbol{C}_{\lambda} \succeq 0, \boldsymbol{C}_{\mu} \succeq 0, \boldsymbol{C}_{\xi} \succeq 0, \\
& \forall j: 1, \cdots, J, t_{10 j} \geq 0, t_{11 j} \geq 0, \\
& \left(t_{10 j}+t_{11 j}\right)-s\left(b_{j}\left(t_{10 j}+t_{11 j}\right)+P_{s} t_{9 j}\right) \geq 0, \\
& \lambda_{j} \geq 0, \mu_{j} \geq 0, \xi_{j} \geq 0, C_{\lambda_{j}} \succeq 0, \boldsymbol{C}_{\mu_{j}} \succeq 0, \boldsymbol{C}_{\xi_{j}} \succeq 0 . \tag{31}
\end{align*}
$$

This is a non-convex optimization problem. We solve this problem using the algorithm described below.

Step1: Find $r_{\text {max }}$ and $s_{\max }$ by solving the following two independent optimization problems:

$$
\begin{align*}
& r_{\max }=\max _{\mathbf{\Phi}^{k 1}, \boldsymbol{\Psi}^{k 2}, t 6, t 7, t 8, \lambda, \mu, \xi, r}  \tag{32}\\
& \text { s.t. } \quad \boldsymbol{\Phi}^{k_{1}} \succeq 0, \boldsymbol{\Psi}^{k_{2}} \succeq 0, t_{6} \geq 0, \\
& \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}(\boldsymbol{v})+N_{0} \boldsymbol{I}\right)\right)+\operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s}, \\
& a\left(t_{7}+t_{8}\right)+P_{s} t_{6}-r\left(t_{7}+t_{8}\right) \geq 0, \\
& \lambda \geq 0, \mu \geq 0, \xi \geq 0, \boldsymbol{C}_{\lambda} \succeq 0, \boldsymbol{C}_{\mu} \succeq 0, \boldsymbol{C}_{\xi} \succeq 0 \text {, }  \tag{33}\\
& \text { and } s_{\text {max }}=\max _{\substack{\boldsymbol{\Phi}^{k 1}, \Psi^{k 2}, t_{9 j}, t_{10 j}, t_{11 j} \\
\lambda_{j}, \mu_{j}, \xi_{j}, \forall j: 1, \ldots, J, s}},  \tag{34}\\
& \text { s.t. } \quad \boldsymbol{\Phi}^{k_{1}} \succeq 0, \quad \boldsymbol{\Psi}^{k_{2}} \succeq 0, \\
& \operatorname{tr}\left(\boldsymbol{\Phi}^{k_{1}}\left(P_{s} \operatorname{diag}(\boldsymbol{v})+N_{0} \boldsymbol{I}\right)\right)+\operatorname{tr}\left(\boldsymbol{\Psi}^{k_{2}}\right) \leq P_{0}-P_{s}, \\
& \forall j: 1, \cdots, J, t_{10 j} \geq 0, t_{11 j} \geq 0, \\
& \left(t_{10 j}+t_{11 j}\right)-s\left(b_{j}\left(t_{10 j}+t_{11 j}\right)+P_{s} t_{9 j}\right) \geq 0, \\
& \lambda_{j} \geq 0, \mu_{j} \geq 0, \quad \xi_{j} \geq 0, \boldsymbol{C}_{\lambda_{j}} \succeq 0, \boldsymbol{C}_{\mu_{j}} \succeq 0, \boldsymbol{C}_{\xi_{j}} \succeq 0 . \tag{35}
\end{align*}
$$

For a given $r$ and $s$, both the problems are convex SDP feasibility problems and both are solved using bisection method. We describe the bisection method in short to solve (32) as follows. Let $r_{\max }$ lie in the interval $\left[r_{l l}, r_{u l}\right]$. Check the feasibility of the constraints of (32) at $r=\left(r_{l l}+r_{u l}\right) / 2$. If feasible then $r_{l l}=r$ else $r_{u l}=r$. Repeat this until $r_{u l}=r_{l l}$ or the desired accuracy is achieved. Maximum values of $r$ and $s$ obtained in the above two independent optimization problems will be larger than the values that would be obtained in the original joint optimization problem (30). This is due to the fact that maximum over a larger set (or unconstrained set) is larger than the maximum over the smaller set (or constrained set). So, the maximum value of the product $r s$ obtained in the constrained optimization problem (30) will be upper bounded by the product $r_{\text {max }} s_{\max }$.

Step2: We represent the optimum value of the optimization problem (30) by $r_{o p t} s_{o p t}$. Having obtained the values of $r_{\max }$ and $s_{\max }$ in Step 1, we obtain $r_{\text {opt }} s_{\text {opt }}$ sequentially by decreasing $r$ from $r_{\text {max }}$ towards zero in discrete steps of size $\Delta_{r}=r_{\max } / N$, where $N$ is a large positive integer, and finding the maximum $s$ such that
constraints in (31) are feasible and the product $r s$ is maximum. The algorithm to obtain $r_{o p t} s_{o p t}$ as follows:

```
for \((\mathrm{i}=\mathrm{N}:-1: 1)\)
begin \(\{\)
\(r_{i}=i * \Delta_{r}\)
\(s_{i}=\quad \max \quad s\)
        \(t_{6}, t_{7}, t_{8}, t_{9 j}, t_{10 j}, t_{11 j}\),
\(\lambda, \mu, \xi, \lambda_{j}, \mu_{j}, \xi_{j}, s, \forall j: 1,2, \cdots, J\)
        subject to all constraints in (31) with \(r=r_{i}\)
if \((i==N)\) then \(r_{o p t}=r_{i}, s_{o p t}=s_{i}\)
elseif \(\left(r_{o p t} s_{o p t} \leq r_{i} s_{i}\right)\) then \(r_{o p t}=r_{i}, s_{o p t}=s_{i}\)
else exit for loop
endif
\(\}\) end for loop
```

For a given value of $r_{i}$ and $s$ in the interval $\left[0, s_{\text {max }}\right]$, the constrained maximization problem in the for loop above is a SDP feasibility problem, and $s_{i}$ can be obtained using the bisection method as described in Step1 to solve (32). With $r_{o p t} s_{o p t}$ from Step 2 above for $\left(k_{1}, k_{2}\right)$ selected relays, worst case secrecy rate $R_{s}^{k_{1} k_{2}}$ for a given source power $P_{s}$ is then given by $R_{s}^{k_{1} k_{2}}=\frac{1}{2} \log _{2} r_{\text {opt }} s_{\text {opt }}$, and the maximum secrecy rate is $R_{s}^{\max }=\max _{\text {all relay combinations }} R_{s}^{k_{1} k_{2}}$. Maximization is performed over all $2^{M}-1$ possible relay combinations.

## IV. Results and Discussions

We evaluated the secrecy rate for AF beamforming with/without cooperative jamming, perfect/imperfect CSI and multiple eavesdroppers for different system scenarios through simulations. The results are generated for $M=2, J=1,2,3$, $N_{0}=1, P_{s}=3 \mathrm{~dB}$ and $N=50$. We take the norm of the CSI error vectors on all links to be equal, and we denote it by $\epsilon$.

In Fig. 2(a), we plot the secrecy rate as a function of total


Fig. 2. Secrecy rate versus total relay transmit power in AF relay beamforming with/without CJ, perfect/imperfect CSI, and multiple eavesdroppers. CJ gives significant gains with 2 and 3 eavesdroppers.
relay transmit power with the following system parameters:


Fig. 3. Secrecy rate versus CSI error ( $\epsilon$ ) in AF relay beamforming with/without CJ, perfect/imperfect CSI and multiple eavesdroppers.
$\sigma_{\gamma_{1}^{*}}=\sigma_{\gamma_{2}^{*}}=4.0, \sigma_{\alpha_{1}^{*}}=4.0, \sigma_{\alpha_{2}^{*}}=0.0, \sigma_{\beta_{11}^{*}}=\sigma_{\beta_{21}^{*}}=4.0$, $\sigma_{\beta_{12}^{*}}=\sigma_{\beta_{22}^{*}}=4.0, \sigma_{\beta_{13}^{*}}=\sigma_{\beta_{23}^{*}}=4.0$, and $\epsilon=0.1$. We assume that there is no direct path from source to destination and source to any eavesdropper. From Figs. 2 (a) and (b), we observe that $i$ ) in the presence of only one eavesdropper $(J=1)$, the advantage of $\mathbf{C J}$ compared to no CJ is not significant, and $i i$ ) with 2 and 3 eavesdroppers $(J=2,3)$, however, significant gains in secrecy rates due to CJ compared to no CJ are achieved. We also observe that imperfect CSI degrades secrecy rates compared to those with perfect CSI, and that increased number of eavesdroppers results in reduced secrecy rates. In Fig. 3, we plot the secrecy rate as a function of CSI error ( $\epsilon$ ) with $P_{0}-P_{s}=6 \mathrm{~dB}$ and remaining system parameters are same as in Fig. 2. From Fig. 3, we observe that i) $\mathbf{C J}$ results in rate gains compared to no $\mathbf{C J}$ for $J=2,3$, and $i i$ ) the achieved rate gain due to CJ is maximum for perfect CSI case and the gain diminishes as the CSI error variance is increased. Next, Fig. 4 shows the secrecy rate


Fig. 4. Secrecy rate versus total relay transmit power in AF relay beamforming with/without CJ, perfect/imperfect CSI, and multiple eavesdroppers. Source-destination and relays-destination channels are stronger than the corresponding eaves channels.
results for a scenario where the rate gains due to CJ is not significant even for $J=2,3$. In this scenario, sourcedestination and relays-destination channels are stronger than the corresponding eaves channels. The corresponding parameters used are: $\sigma_{\gamma_{1}^{*}}=\sigma_{\gamma_{2}^{*}}=4.0, \sigma_{\alpha_{0}^{*}}=2.0, \sigma_{\alpha_{1}^{*}}=\sigma_{\alpha_{2}^{*}}=4.0$,
$\sigma_{\beta_{01}^{*}}=0.5, \sigma_{\beta_{02}^{*}}=1.0, \sigma_{\beta_{03}^{*}}=1.5, \sigma_{\beta_{11}^{*}}=\sigma_{\beta_{21}^{*}}=1.0$,
$\sigma_{\beta_{12}^{*}}=\sigma_{\beta_{22}^{*}}=2.0, \sigma_{\beta_{13}^{*}}=\sigma_{\beta_{23}^{*}}=3.0$, and $\epsilon=0.1$. In summary, the secrecy rate gains due to CJ depends on the channel/noise conditions, number of eavesdroppers, and CSI error variances, and the proposed solution allows us to compute the secrecy rate in AF beamforming with CJ under various channel conditions and scenarios.

## V. Conclusions

We evaluated the worst case secrecy rates in AF relay beamforming scheme with cooperative jamming and in the presence of imperfect CSI (using a norm-bounded CSI error model) and multiple eavesdroppers, where the number of eavesdroppers can be more than the number of relays. We solved the optimization problem to find the optimum relay beamforming weights (weights of both data relays and jamming relays) subject to a total relay power constraint and CSI error constraints, and computed the worst case secrecy rate by relaxing the rank one constraint on the complex semidefinite data relays and jamming relays weight matrices and reformulated the optimization problem into a form that was solved using convex semi-definite programming.

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