Secure Communications via Power Control in a MIMO Wiretap Interference Network with Jamming Transmitters and Receivers

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Abstract

We study secure and distributed power control in a multi-link interference network that is tapped by an external eavesdropper. To conceal information from the eavesdropper, legitimate links are equipped with both transmit-based friendly jamming (TxFJ) and receiver-based friendly jamming (RxFJ). Each transmitter-receiver (Tx-Rx) pair seeks to maximize its secrecy rate by determining the best power assignment (PA) for the information, TxFJ, and RxFJ signals. The joint optimization of these parameters is a non-convex problem, thus computationally demanding. Hence, each link should seek for suboptimal solutions. To do so, we aim to provide positive secrecy for each link. Despite its sub-optimality, such an approach precludes the possibility of employing a strong multiuser detector by the eavesdropper (such as successive interference cancellation). We show that a careful assignment of TxFJ and RxFJ powers of a link can provide it with a positive secrecy rate. The TxFJ PA at a link is done with respect to the observed interference at the corresponding Rx, whereas the RxFJ of that link is adjusted using an on-off PA that depends only on the link’s local channel state information (CSI). Hence, the RxFJ adjustment is unaffected by interference fluctuations at the eavesdropper, which facilitates our distributed design. With every link following such a strategy, we model this interaction as a non-cooperative game. Assuming knowledge of eavesdropper’s CSI (E-CSI), we derive sufficient conditions for the uniqueness of the resulting Nash equilibrium. We then propose an algorithm to implement the PA game. Lastly, we relax

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knowledge of E-CSI and propose a framework that is robust to unknown E-CSI at links. Our results indicate that the robust framework has a performance close to when E-CSI is fully known to legitimate links. Moreover, it is shown empirically that the secrecy sum-rate scales with the power budget of legitimate transmitters.

**Index Terms**

Interference network, friendly jamming, full-duplex radios, game theory, distributed design

**I. INTRODUCTION**

Physical-layer (PHY-layer) security has recently gained considerable attention because of its potential to provide secrecy at low computational overhead. This makes such an approach particularly suitable for applications where it is either expensive or computationally demanding to use cryptographic methods. The most common scenario for information-theoretic PHY-layer security is the so-called *wiretap channel*. The wiretap channel involves communications between a legitimate transmitter (Alice) and a corresponding receiver (Bob); such communications is to be secured from an eavesdropper (Eve).

Among proposed methods for PHY-layer security, artificial noise (or friendly jamming) has been noticeably the subject of many research efforts. According to this method [2], Alice can use multiple antennas and a portion of her transmit power to create a bogus signal –known as *artificial noise* or *transmit-based friendly jamming* (TxFJ)— alongside the information signal to confuse a nearby Eve. Assuming that Alice knows Alice-Bob channel, she creates TxFJ via precoding techniques such that the precoded TxFJ signal falls in the null-space of Alice-Bob channel, hence not affecting Bob’s reception. Alongside the TxFJ method, secrecy can also be provided with the help of another node (e.g., a relay) that is dedicated to generate friendly jamming (FJ) signals [2]. Such a method is usually referred to as *cooperative jamming (CJ)*\(^1\). Despite having the same effect as the TxFJ method, the CJ approaches face several implementation issues compared to the TxFJ method such as mobility, trustworthiness and synchronization.

To address the issues related to the CJ approach, it has been suggested in [3] to equip Bob with in-band full-duplex (FD) capabilities, allowing him to generate his own friendly jamming signal while receiving the information signal from Alice. Such an FJ signal is hereafter referred to as *Rx-based friendly jamming* (RxFJ) [3]. Hence, using RxFJ, the disadvantages of CJ can

\(^1\)Same as TxFJ, the FJ signals emitted from the helper node in CJ do not affect Bob’s reception.
be mitigated, and still a node exists in the system which is dedicated for generating FJ signals (i.e., Bob). Another example of using RxFJ is [4], where the authors proposed an optimal power allocation framework to maximize the secrecy rate of a link that is tapped by a single-/multiple-antenna Eve under different configurations (e.g., deterministic/statistical knowledge of channel between Alice and Eve, multi-antenna Bob/Eve). Other works consider PHY-layer security when FD capability is adopted at both Alice and Bob for bidirectional communications, i.e., Bob transmits information signals to Alice rather than generating RxFJ (see [5] and its references). Our focus in this paper is the case where Bob is employed to generate RxFJ.

While the single-link scenario is of great importance in developing early observations, secrecy analysis for multi-link settings introduces new challenges not present in the single-link scenario. The definition of secrecy in multi-link settings depends on the specific network under consideration. For instance, links might be interested in transmissions of their neighboring links. Thus, the design must ensure that a given link’s transmission is secured from other links. Such a network is referred to as multi-link channel with confidential messages. Another possibility is when external Eves exist in the network and the transmissions of legitimate links must be kept secure from these Eves. Such a network is referred to as multi-link wiretap channel.

In this paper, we study PHY-layer security in a multi-link wiretap channel. In our network model, legitimate links share the same bandwidth for their transmissions, thus interfering with one another; at the same time, an eavesdropper snoops on ongoing communications, hence the name wiretap interference channel. Legitimate links are capable of TxFJ and RxFJ. Our design parameters are the power of RxFJ, and the power assignment (PA) between the information and TxFJ signals. Our work is motivated by the following simple observation: For a given link, when no secrecy is required, the higher the Alice’s power budget, the higher the information rate at the respective Rx. However, when secrecy is also a requirement, although the information rate still increases monotonically with the power at Alice, the secrecy rate may not be necessarily so because more power transmitted from Alice also increases the leakage rate at Eve.

Our first objective is to find a setting under which the secrecy rate of a link increases monotonically with Alice’s power. We find a lower bound on the power allocated to TxFJ above which positive secrecy is achievable for a given link. Once positive secrecy is achieved, the secrecy rate becomes a monotonically increasing function of power at Alice, thus having the same trend as the information rate. Therefore, the rest of Alice’s power is allocated to the information signal. Clearly, having a larger power budget at Alice leads to larger secrecy
rates. Although guaranteeing positive secrecy does not offer any sort of optimality in terms of individual or network-wide secrecy, it ensures that no link experiences zero secrecy. In contrast, when the aim is to maximize the sum of secrecy rates, we cannot ensure that every link achieves a non-zero secrecy rate [6]. A zero secrecy scenario can be exploited by Eve, who can perform sophisticated multiuser detection techniques (e.g., successive interference cancellation or SIC) to decode ongoing communications. Such an issue was reported in [7], and it was shown in [8] that an SIC-capable Eve can significantly decrease the network secrecy if some links experience zero secrecy rates. Provided that all links have non-zero secrecy rates, Eve cannot apply SIC.

In our framework, the lower bound on the TxFJ power of an Alice (which guarantees positive secrecy) depends on the amount of interference at her corresponding Bob. As for the RxFJ power, though its setting generally depends on multiuser interference (MUI), we show that to preserve positive secrecy, it is sufficient for each Bob to set RxFJ based on only his local channels (i.e., Alice-Bob and Bob-Eve channels). This result offers a great simplification to our distributed design, as Bobs do not require to adjust RxFJ w.r.t. MUI which is often varying in time.

We assume that when setting their transmission parameters, there is no centralized authority responsible for computations and optimization. Hence, links have to make decisions in a distributed fashion. Such a design inevitably produces interference at several links. However, because Eve also receives interference from all links, a careful design ensures that interference at legitimate links is properly managed while interference at Eve is kept high as much as possible. We model such an interaction between legitimate links using the theory of non-cooperative games.

The works in [9], [10] studied secure precoding in wiretap interference networks. Moreover, the authors in [11] studied power control in a multi-channel interference network without considering TxFJ and RxFJ. All of these works assumed full knowledge of the eavesdropper’s channel state information (E-CSI) at each Alice, which may not be a practical assumption. In addition, to the best of our knowledge, the solutions proposed in those works are not extendable to cover the case of unknown E-CSI. Regarding the PA between the information and TxFJ signals, the works in [12] and [13] focused only on a single-link scenario, and their approaches are not extendable to the case of multiple links. The authors of [14] exploited full-duplex capability in the base station of a broadcast/multiple-access wiretap channel to secure multiple half-duplex downlink

\footnote{A full description of the effect of a zero secrecy rate on the secrecy of an interference network was given in [8], where we showed that Eve can cancel the interference coming from links with zero secrecy rates, thus increasing the signal-to-interference-plus-noise-ratio (SINR) while snooping on other transmissions with non-zero secrecy rates.}
and uplink users by generating RxFJ/TxFJ for uplink/downlink communications. They proposed a multi-objective optimization framework to find the best trade-off in minimizing downlink and uplink powers consumed at the base station (BS), subject to given constraints on information and secrecy rates of downlink and uplink users. The work in [15] studied power minimization for the information, TxFJ and RxFJ signals in a broadcast channel with confidential messages under given guarantees on the individual secrecy rate for each Bob. In [15], the BS is responsible to do the optimization. This indicates that the power minimization is done in centralized fashion, for which the BS must acquire the CSI between itself and all downlink users. We investigate a more challenging scenario (i.e., interference channel) where contrary to [15], distributed computation and limited coordinations are required [16]. Overall, our contributions can be summarized as follows:

- Using TxFJ and RxFJ, we define a lower bound on the power allocated to the TxFJ that guarantees positive secrecy of transmissions at each given link.
- We propose a non-cooperative game to model the power control problem in the interference network under our study. Under the assumption that Alice-/Bob-Eve channels are fully known, we derive sufficient conditions under which the proposed non-cooperative game admits a unique Nash equilibrium (NE).
- We propose alternative sufficient conditions for the uniqueness of the NE. Such conditions allow for predicting the existence of a unique NE in distributed fashion.
- We show that our distributed design can be implemented using an asynchronous update algorithm. This algorithm is robust to transmission delays over various links.
- Lastly, we relax the assumption of full knowledge of E-CSI at each Alice and propose a version of our algorithm that is robust to uncertainties in knowledge of E-CSI.

We need to emphasize that in this paper, we first examine our proposed distributed design under full knowledge of E-CSI. Although availability of E-CSI at all links is not a practical assumption, we can build foundations for our distributed algorithm to easily introduce important performance metrics. Once, our analysis is complete, we relax knowledge of E-CSI and propose the version of our algorithm that is robust to uncertainties in E-CSI knowledge.

**Notation:** Boldface uppercase/lowercase letters denote matrices/vectors. $a \geq b$ denotes the element-wise inequality between vectors $a$ and $b$. The matrix $\mathbf{I}$ is the identity matrix of appropriate size. $E[\bullet]$, $\bullet^\dagger$, and $\text{Tr}(\bullet)$ are, respectively, the expected value, complex conjugation (with
transposition in case of vectors and matrices), and the trace of a matrix. The sets of real and complex numbers are indicated by \( \mathbb{R} \) and \( \mathbb{C} \), respectively.

II. System Model

We first describe a model for the network under consideration and introduce the main performance metrics. Consider \( Q \) transmitters (\( Q \geq 2 \)), Alice, ..., Alice_\( Q \) that communicate with their respective receivers, Bob_\( 1 \),..., Bob_\( Q \). Let \( Q \triangleq \{ 1, 2, \ldots, Q \} \). Alice_\( q \), \( q \in Q \), has \( N_q \) transmit antennas, and Bob_\( q \) has \( M_q \) antennas. A passive Eve with \( L \) antennas is also present in the communication range\(^3\). The received signal at Bob_\( q \) is

\[
y_q = \bar{H}_{qq}u_q + \sqrt{\tau_q}H'_{qq}\mathbf{m}_q + \sum_{r=1, r\neq q}^{Q} (\bar{H}_{rq}u_r + H'_{rq}\mathbf{m}_r) + \mathbf{n}_q
\]

where \( \bar{H}_{rq} \in \mathbb{C}^{M_q \times N_r}, r \in Q \), is the \( M_q \)-by-\( N_r \) complex channel matrix between Alice_\( r \) and Bob_\( q \), \( \mathbf{u}_q \in \mathbb{C}^{N_q} \) is the transmitted signal from Alice_\( q \), \( \tau_q \in \mathbb{R}^+ \), and \( H'_{qq} \in \mathbb{C}^{M_q \times M_q} \) are, respectively, the positive-real-valued self-interference-suppression (SIS) factor and the self interference channel at Bob_\( q \) due to the imperfect SIS at Bob_\( q \).\(^4\) Such a model for self-interference was adopted in several recent works (see \([14] \), \([18] \)), and practical implementations of it exist in the literature (see e.g., \([19] \)). \(^5\) \( \mathbf{m}_q \in \mathbb{C}^{M_q} \) is the RxFJ signal created by Bob_\( q \), which is a zero mean circularly symmetric complex Gaussian random variable (ZMCSCG-RV) with covariance matrix of \( E[\mathbf{m}_q\mathbf{m}_q^\dagger] = p'_q \mathbf{I} \) where \( \text{Tr}(\mathbf{m}_q\mathbf{m}_q^\dagger) = M_qp'_q \leq P'_q \) and \( P'_q \) denotes the total power of Bob_\( q \) to be used for RxFJ. \( H'_{rq} \in \mathbb{C}^{M_q \times M_r}, r \neq q \), is the channel from Bob_\( r \) to Bob_\( q \) because the RxFJ created by other Bobs interfere with Bob_\( q \)'s reception. \( \mathbf{n}_q \in \mathbb{C}^{M_q} \) is the complex additive white Gaussian noise (AWGN) whose covariance matrix is \( E[\mathbf{n}_q\mathbf{n}_q^\dagger] = N_0 \mathbf{I} \) with \( N_0 \in \mathbb{R}^+ \). We assume that \( \bar{H}_{rq} = \bar{H}_{rq}d_{rq}^{-\eta/2} \) where \( \bar{H}_{rq} \in \mathbb{C}^{M_q \times N_r} \) represents the small-scale fading, \( d_{rq} \) is the distance between Alice_\( r \) and Bob_\( q \) in meters and \( \eta \) is the path-loss exponent. The same equivalent assumption holds for \( H'_{rq}, r \neq q \), i.e., \( H'_{rq} = \bar{H}'_{rq}d_{rq}^{-\eta/2} \) where \( \bar{H}'_{rq} \in \mathbb{C}^{M_q \times M_r} \) and \( d_{rq} \) is the distance from Bob_\( r \) to Bob_\( q \).

\(^3\)Note that \( L \) can be assumed to be large enough to represent multiple multi-antenna colluding eavesdroppers \([2] \). However, in this paper, for ease of presentation, we consider the \( L \)-antenna Eve as a single entity.

\(^4\)Recall that enabling in-band full-duplex communications requires suppression of the transmitted signal of the FD-enabled device at its receive chain to allow for simultaneous transmission and reception. However, such suppression may not be perfect, and there is often a residual self-interference captured at the receive chain \([17] \).

\(^5\)We assume that FD receivers are not experiencing dynamic range issues (as pointed out in \([20] \)), which causes the additive noise at the receive chain to be dependent of the transmit power of the FD device. Incorporating this assumption can be considered as a subject for future research.
The received signal at Eve is

\[ \mathbf{z} = \tilde{\mathbf{G}}_q \mathbf{u}_q + \mathbf{G}'_q \mathbf{m}_q + \sum_{r=1}^{Q} (\tilde{\mathbf{G}}_r \mathbf{u}_r + \mathbf{G}'_r \mathbf{m}_r) + \mathbf{e} \]  

(2)

where \( \tilde{\mathbf{G}}_q \in \mathbb{C}^{L \times N_q} \), \( q \in \mathcal{Q} \) denotes the complex channel matrix between Alice_\( q \) and Eve. Let \( \tilde{\mathbf{G}}_q = \overline{\mathbf{G}}_q d^{-\eta/2} \) where \( \overline{\mathbf{G}}_q \in \mathbb{C}^{L \times N_q} \) and \( d_{qe} \) is the distance between Alice_\( q \) and Eve. \( \mathbf{G}'_q \in \mathbb{C}^{L \times M_q} \), \( q \in \mathcal{Q} \), is the channel between Bob_\( q \) and Eve, and \( \mathbf{G}'_q = \overline{\mathbf{G}}'_q d'^{-\eta/2} \) where \( \overline{\mathbf{G}}'_q \in \mathbb{C}^{L \times M_q} \) and \( d'^{qe} \) is the distance from Bob_\( q \) to Eve. Finally, \( \mathbf{e} \) has the same characteristics as \( \mathbf{n}_q \). The signal \( \mathbf{u}_q = \mathbf{s}_q + \mathbf{w}_q \) consists of the information signal \( \mathbf{s}_q \) and TxFJ \( \mathbf{w}_q \). We only consider the case of single-stream data transmission using multiple antennas. That is, we set \( \mathbf{s}_q \triangleq \mathbf{T}_q x_q \) where \( \mathbf{T}_q \in \mathbb{C}^{N_q} \) is the precoder and \( x_q \in \mathbb{C} \) is the information signal. In fact, we use multiple transmit and receive antennas at each link to exploit the maximum diversity of the MIMO link, and not exploit the spatial multiplexing, i.e., multiple antennas are exploited for beamforming\(^6\). Although in our scenario, beamforming is a suboptimal approach, it helps us to gain valuable insight into solving the underlying optimization problems. Beamforming has been shown to be optimal under several channel models (see [21]).

Assume that a Gaussian codebook is used for \( x_q \), i.e., \( x_q \) is distributed as a ZMCS-CG-RV with \( E[x_q x_q^\dagger] = \phi_q P_q \) where \( P_q \) is the total transmit power of Alice_\( q \) and \( 0 \leq \phi_q \leq 1 \) is the portion of transmit power allocated to the information signal. For the TxFJ, we write \( \mathbf{w}_q \triangleq \mathbf{Z}_q \mathbf{v}_q \), where \( \mathbf{Z}_q \in \mathbb{C}^{N_q \times (N_q - 1)} \) is the precoder for the TxFJ signal and \( \mathbf{v}_q \in \mathbb{C}^{(N_q - 1)} \) is the TxFJ signal with i.i.d. ZMCS-CG entries and \( E[\mathbf{v}_q \mathbf{v}_q^\dagger] = \sigma_q \mathbf{I} \). The scalar value \( \sigma_q = \frac{(1-\phi_q)P_q}{N_q-1} \) denotes the TxFJ power\(^7\). Let \( \tilde{\mathbf{H}}_{qq} = \mathbf{U}_q \Sigma_q \mathbf{V}_q^\dagger \) denote the singular value decomposition (SVD) of \( \tilde{\mathbf{H}}_{qq} \) where \( \Sigma_q \) is the diagonal matrix of singular values in descending order, and \( \mathbf{U}_q \) and \( \mathbf{V}_q \) are left and right matrices of singular vectors, respectively. We set \( \mathbf{Z}_q = \mathbf{V}_q^{(2)} \) where \( \mathbf{V}_q^{(2)} \) is the matrix of \( N_q - 1 \) rightmost columns of \( \mathbf{V}_q \) corresponding to the smallest singular values [2]. We assume that Alice_\( q \) knows the channel \( \tilde{\mathbf{H}}_{qq} \)\(^8\). The precoder \( \mathbf{T}_q \) is set to \( \mathbf{T}_q = \mathbf{V}_q^{(1)} \), where \( \mathbf{V}_q^{(1)} \) is the

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\(^6\)Later on, we explain the rationale behind choosing beamforming over spatial multiplexing.

\(^7\)Notice that the TxFJ power is distributed uniformly between different dimensions of \( \mathbf{v}_q \). In the case of full knowledge of E-CSI, such power division is not optimal. However, when no knowledge of E-CSI is available (which we assume later in this paper), it was shown that uniform division of TxFJ power between different dimensions of \( \mathbf{v}_q \) is the optimal approach (see [2], [12]).

\(^8\)Acquiring channel state information (CSI) between Alice_\( q \) and its corresponding Bob_\( q \) is assumed to be done securely. For example, a two-phase channel estimation can be performed, where in the first/second time-slot, Alice_\( q \)/Bob_\( q \) sends the pilot signals to Bob_\( q \)/Alice_\( q \). This way, we avoid having to send explicit CSI feedback from one communication end to another, thus lowering the probability of eavesdropping on channel estimates.
first column of \( V_q \) corresponding the largest singular value, to execute transmit beamforming [22]. Let \( H_{qq} \triangleq \tilde{H}_{qq} V_q^{(1)} \), \( H_{jqq} \triangleq \tilde{H}_{jqq} V_q^{(2)} \), \( H_{q} \triangleq \tilde{H}_{q} V_q^{(1)} \), \( H_{jq} \triangleq \tilde{H}_{jq} V_q^{(2)} \), \( G_q \triangleq \tilde{G}_q V_q^{(1)} \), \( G_{jq} \triangleq \tilde{G}_q V_q^{(2)} \). The terms \( G_q \) and \( G_{jq} \), \( \forall q \in Q \) denote the E-CSI components. Hence, (1) and (2) can be written as

\[
\begin{align*}
y_q &= H_{qq}x_q + H_{jqq}v_q + \sqrt{\tau_q}H_{qq}^\prime m_q + \sum_{r=1}^{Q} (H_{rq}x_r + H_{jrq}v_r + H_{rq}^\prime m_r) + n_q \quad \text{(3a)} \\
z &= G_q x_q + G_{jq} v_q + G_{q}^\prime m_q + \sum_{r=1}^{Q} (G_q x_r + G_{jq} v_r + G_{jq}^\prime m_r) + e. \quad \text{(3b)}
\end{align*}
\]

An illustration of the system model under study is given in Fig. 1 for a two-link network. It can be seen that the interference components at each Bob include his self-interference signal as well as information, TxFJ and RxFJ signals of unintended links. Eve also receives all information, TxFJ and RxFJ signals.

After receiving \( y_q \) at Bob, a linear receiver \( d_q \in \mathbb{C}^{M_q} \) is applied. Given that \( d_q^H H_{jq} v_q = 0^9 \),

\(^9\text{Note that the choice of the linear receiver (to be discussed near the end of this section) affects this assumption. In this paper, we choose the linear receiver such that this assumption holds.}\)
the linear estimate of $x_q$ at Bob is

$$
\hat{x}_q = d_q^\dagger(\mathbf{H}_{qq}x_q + \sqrt{\tau_q}\mathbf{H}'_{qq}\mathbf{m}_q + \sum_{r=1 \atop r \neq q}^Q (\mathbf{H}_{rq}x_r + \mathbf{H}_{jq}v_r + \mathbf{H}'_{rq}\mathbf{m}_r) + \mathbf{n}_q).
$$

Hence, the information rate for the $q$th link can be expressed as\(^{10}\)

$$
C_q = \log(1 + \frac{\phi_q P_q}{a_q + b_q p'_q})
$$

where

$$
a_q = \frac{\sum_{r=1 \atop r \neq q}^Q \left( |d_q^\dagger \mathbf{H}_{rq}|^2 \phi_r P_r + |d_q^\dagger \mathbf{H}_{jq}|^2 \sigma_r + |d_q^\dagger \mathbf{H}'_{rq}|^2 p'_r \right) + N_0}{|d_q^\dagger \mathbf{H}_{qq}|^2}
$$

$$
b_q = \frac{\tau_q |d_q^\dagger \mathbf{H}'_{qq}|^2}{|d_q^\dagger \mathbf{H}_{qq}|^2}.
$$

Eve also applies the linear receiver $\mathbf{r}_q \in \mathbb{C}^L$ while eavesdropping on $q$th link’s signal to obtain the following estimate on $x_q$

$$
\hat{z}_q = \mathbf{r}_q^\dagger(\mathbf{G}_q x_q + \mathbf{G}_{jq} v_q + \mathbf{G}'_q \mathbf{m}_q + \sum_{r=1 \atop r \neq q}^Q (\mathbf{G}_r x_r + \mathbf{G}_{jr} v_r + \mathbf{G}'_r \mathbf{m}_r) + \mathbf{e}).
$$

Thus, the rate at Eve while eavesdropping on Alice$_q$ (i.e., leaked rate of Alice$_q$ at Eve) is

$$
C_{eq} = \log(1 + \frac{\phi_q P_q}{c_q + d_q p'_q})
$$

\(^{10}\)Note that the distribution characteristics of the additive noise do not change after passing through the linear receiver [22].
where
\begin{align}
c_q &= \frac{|r_q^i G_{jq}|^2 \sigma_q}{|r_q^i G_q|^2} + \\
\sum_{r \neq q}^Q &\left( |r_q^i G_r|^2 \sigma_r + |r_q^i G_{jr}|^2 + |r_q^i G'_{r'}|^2 p'_r \right) + N_0 \\
d_q &= \frac{|r_q^i G'_q|^2}{|r_q^i G_q|^2}.
\end{align}

Finally, the secrecy rate of Alice, can be written as
\begin{equation}
C_{q}^\text{sec} = \max\{C_q - C_{eq}, 0\}.
\end{equation}

The linear receivers \(d_q\) and \(r_q\), \(q \in Q\), are assumed to be chosen according maximal ratio combining (MRC) method to maximize the reception of signal at Bob, and Eve, respectively. Hence, \(d_q = U_q^{(1)}\) where \(U_q^{(1)}\) is the first column of \(U_q\) (recall that \(\tilde{H}_{qq} = U_q \Sigma_q V_q^\dagger\)). Using this linear receiver, the TxFJ signal of Alice will be nullified at Bob as shown in the system model in Fig. 1, i.e., Bob does not receive the TxFJ from Alice, \(q \in Q\). In other words, \(d_q^\dagger H_{jq} v_q = 0, q \in Q\) (see (4)). Let the SVD of \(\tilde{G}_q\) be denoted as \(\tilde{G}_q = L_q D_q R_q\) where \(L_q\) and \(R_q\) are matrices of left and right singular vectors, respectively, and \(D_q\) is the diagonal matrix of singular values in descending order. Thus, while eavesdropping on \(q\)th link, Eve sets its linear receiver \(r_q = L_q^{(1)}\) where \(L_q^{(1)}\) is the first column of matrix \(L_q\).

We need to emphasize that the choice of precoder (i.e., beamformers) for TxFJ signal in this paper is mainly driven by the fact that acquiring E-CSI knowledge may not be possible in the cases where eavesdropper is a passive node. For a single-link scenario, it was shown in [24] that optimizing the precoders of information and TxFJ signals requires complete knowledge of E-CSI. However, in this paper, the beamforming vector of TxFJ signal for each link depends only on the channel between the two nodes comprising that link, which is relatively more practical to achieve.

\[\text{Note that in specifying the secrecy rate of a link, because no link has the knowledge on whose transmission Eve is interested, all legitimate links protect their transmissions from Eve. Thus, the secrecy rate of each link can be determined by (10) (see [23]).}\]

\[\text{Note that other decoders (such as MMSE [22]) can also be employed by Eve. This issue will be discussed later in the simulation section.}\]
Our choice of beamforming vector of information signal for each link comes from the fact that
the number of antennas at Eve may not be known in some cases. As pointed out in [2], the main
limitation of the TxFJ method is that if Eve has more antennas than Alice, then Eve may be able
to nullify the effect of TxFJ on itself by a specific choice of decoder (i.e., linear receiver) at its
receive antennas. Looking at (7), the general form of the TxFJ signal from the Alice
\( q \) that is
received at Eve can be written as
\[
\mathbf{r}^\dagger_q \tilde{\mathbf{G}}_q \mathbf{V}_q \mathbf{v}_q
\]
where \( \mathbf{V}_q \) is the \( N \) rightmost columns of \( \mathbf{V}_q \), \( 0 \leq N \leq N_q - 1 \). Let \( \tilde{\mathbf{G}}_q \mathbf{V}_q = \mathbf{L}'_q \mathbf{D}'_q \mathbf{R}'_q \) be the SVD of the \( L \times N \) matrix \( \tilde{\mathbf{G}}_q \mathbf{V}_q \), where
\( \mathbf{L}'_q \) and \( \mathbf{R}'_q \) are matrices of left and right singular vectors, respectively, and \( \mathbf{D}'_q \) is the diagonal
matrix of singular values. If \( L > N \), indicating that \( \tilde{\mathbf{G}}_q \mathbf{V}_q \) is a tall matrix, then Eve has more
antennas than the total dimensions considered for the TxFJ signal of Alice \( q \). Hence, if Eve knows
\( \tilde{\mathbf{G}}_q \mathbf{V}_q \), she can choose \( \mathbf{r}_q \) to be the rightmost \( L - N \) columns of the matrix \( \mathbf{L}'_q \). This way, Eve
can nullify the TxFJ signal, i.e.,
\[
\mathbf{r}^\dagger_q \tilde{\mathbf{G}}_q \mathbf{V}_q \mathbf{v}_q = 0
\]
Therefore, with sufficiently high number of antennas, Eve can nullify the effect of TxFJ on herself. To prevent this, we need to make sure
that \( L - N \leq 0 \), so \( N \geq L \). To ensure that \( N \geq L \), Alice \( q \) Tx uses as many dimensions for the
TxFJ signal as possible. Hence, we set \( N \) to its maximum value, i.e., \( N = N_q - 1 \), meaning that
\( \mathbf{V}_q \) is the \( N_q - 1 \) rightmost columns of \( \mathbf{V}_q \), so \( \mathbf{V}'_q = \mathbf{V}_q^{(2)} \) (see also (3) and the description above
it). This way, at least we know that Alice \( q \) cannot do any better to prevent nullification of TxFJ
on Eve. Obviously, by choosing, for example, \( N = 1 \) (i.e., allocating one dimension to the TxFJ
precoder), even an Eve with \( L = 2 \) antennas can nullify the effect of TxFJ on herself.

As mentioned before, the information signal \( \mathbf{s}_q \) can be written as \( \mathbf{s}_q = \mathbf{T}_q \mathbf{x}_q \), where \( \mathbf{T}_q \) is the
precoding matrix (precoder) and \( \mathbf{x}_q \) is the information signal. With the aforementioned choice of
TxFJ beamformer, the beamformer that can maximize the information rate of the Alice \( q \) would
be \( \mathbf{T}_q = \mathbf{V}_q^{(1)} \), where \( \mathbf{V}_q^{(1)} \) is the \( N_q - N = N_q - (N_q - 1) = 1 \) leftmost column of \( \mathbf{V}_q \), i.e., the
first column of \( \mathbf{V}_q \). Such a choice of precoders forces \( \mathbf{x}_q \) to be a scalar value, signifying that
only single-stream signals are allowed to be transmitted.

Overall, with these choices of precoders, we first make sure that our precoders do not require
knowledge of E-CSI, then we make sure that our TxFJ signal will not be nullified at an Eve
that has relatively low number of antennas. Such an approach in assigning precoders was also
used in [25], [26]. Notice that with knowledge of number of antennas at Eve, the exact amount
of dimensions for the TxFJ beamformer can be chosen to ensure that Eve is not able to nullify
the TxFJ at herself. However, in the case of collusion between multiple eavesdroppers, they can
form a MIMO receiver with higher number of receive antennas.
III. PROBLEM FORMULATION

In this section, we present necessary bounds to achieve positive secrecy and make foundations for our game-theoretic formulation. We form the following optimization problem for $q \in Q$:

$$\begin{align*}
\text{maximize} & \quad C_q^{sec} \\
\text{s.t.} & \quad 0 \leq \phi_q \leq 1 \\
& \quad 0 \leq p'_q \leq P'_q.
\end{align*}$$

(11)

Due to the non-concavity of the objective function in (11) w.r.t. decision variables, the optimization in (11) is non-convex\(^{13}\). To find a tractable (and yet suboptimal) solution, we decompose the analysis of RxFJ and PA into two sub-problems. We first propose a tractable solution for $p'_q$ that results in not only maintaining positive secrecy, but also alleviating the need for knowledge of MUI at Eve. Then, we propose a method to find a suboptimal value for the PA between information and TxFJ signals.

A. RxFJ Power Assignment

Removing the $\max\{\cdot\}$ and $\log(\cdot)$ operators from $C_q^{sec}$ in (10), the secrecy maximization w.r.t. $p'_q$ can be written as

$$\begin{align*}
\text{maximize} & \quad 1 + \frac{\phi_q P_q}{a_q + b_q P_q} \\
\text{s.t.} & \quad 0 \leq p'_q \leq P'_q.
\end{align*}$$

(12)

One can do a simple one-dimensional search to find the optimal value of $p'_q$. However, such an approach demands knowledge of MUI at Eve to be available, which may be difficult to achieve. In the remainder of this section, we propose a method for setting RxFJ power that can set the RxFJ PA independent of MUI at Eve.

We first find conditions that result in having positive secrecy at link $q$. Positive secrecy in (10) is achievable if and only if the objective value in (12) is larger than one. Furthermore, it

\(^{13}\)The non-concavity of objective function can be easily seen by examining the Hessian matrix of the objective function.
can be easily shown that the objective value in (12) is larger than one if and only if the optimal objective value of the following optimization is larger than one:

\[
\max_{p'_q} \quad g(p'_q) \triangleq \frac{\phi_q P_q}{a_q + b_q p'_q} = \frac{c_q + d_q p'_q}{a_q + b_q p'_q}
\]

s.t. \( 0 \leq p'_q \leq P'_q. \) (13)

Note that the relationship between the solutions of (12) and (13) (that result in their corresponding objective values being larger than one) is of necessary and sufficient type. Hence, if we are seeking for a set of conditions/solutions that result in positive secrecy, we can examine those solutions by judging on the objective value that they yield for (13) instead of (12). The first and second derivatives of \( g(p'_q) \) are as follows

\[
\frac{dg(p'_q)}{dp'_q} = \frac{b_q c_q - a_q d_q}{(a_q + b_q p'_q)^2} \quad \text{(14a)}
\]

\[
\frac{d^2g(p'_q)}{dp'_q^2} = 2b_q \frac{b_q c_q - a_q d_q}{(a + bp')^3}. \quad \text{(14b)}
\]

Hence, the optimal value of \( p'_q \) (i.e., \( p'_q^* \)) that solves (13) is as follows:

\[
p'_q^* = \begin{cases} 
P'_q & \text{if } b_q < \frac{a_q d_q}{c_q} \\
0 & \text{if } b_q > \frac{a_q d_q}{c_q}.
\end{cases} \quad \text{(15)}
\]

Simplifying the first condition of (15), a threshold for SIS factor is as follows

\[
\tau_q < \frac{|d^\dagger_q H_{qq}|^2 a_q d_q}{|d^\dagger_q H_{qq'}|^2 c_q}. \quad \text{(16)}
\]

Later on, we show in simulations that whenever positive secrecy is achievable (i.e., the objective in (12) is larger than one), the optimal value of RxFJ can be frequently derived from (15), signifying that the solution to (13) is very likely the optimal solution to (12) as well (see Fig. 2).

Considering (16), we can conclude the following: Given \( c_q \) and \( d_q \), if the (normalized) MUI at Bob_q (\( a_q \)) is not as strong as the (normalized) self-interference channel \( \frac{|d^\dagger_q H_{qq}|^2}{|d^\dagger_q H_{qq'}|^2} \), i.e., if \( \frac{|d^\dagger_q H_{qq}|^2 a_q}{|d^\dagger_q H_{qq'}|^2} \)

14 One can simply set the objective of (12) to be larger than one and end up with \( g(p'_q) > 1 \) (and vice versa) where \( g(p'_q) \) is defined in (13).

15 Although when \( p'_q = 0 \) the benefits of RxFJ are not present, one can set a minimum RxFJ power to prevent RxFJ from going to zero.
is small, the power of RxFJ should be very weak to maintain positive secrecy, leading to \( p'_q^* = 0 \). However, if \( \frac{|d_q H_{qq}|^2}{|d_q H_{qq}|^2} \) is large, the effect of RxFJ on Bob_q is not as significant as MUI, so less suppression of self-interference can be allowed and still maintain positive secrecy, i.e., \( p'_q^* = P'_q \) becomes the favorable solution. An equivalent intuition holds for \( d_q/c_q \) when \( \frac{|d_q H_{qq}|^2}{|d_q H_{qq}|^2} \) and \( a_q \) are given. Specifically, a large \( d_q/c_q \) indicates that RxFJ degrades Eve’s reception more than the MUI \( (c_q) \) received at Eve. Hence, the FD-enabled Bob_q can perform less suppression, i.e., \( p'_q^* = P'_q \) becomes the favorable solution.

It can be seen in (15) that the optimal value of RxFJ that solves (13) depends on two factors: MUI at Bob_q (i.e., \( a_q \)) and MUI at Eve while eavesdropping the \( q \)th link (i.e., \( c_q \)), which are defined in (6) and (9), respectively. It may not be practical for a legitimate node to have knowledge on MUI at Eve because Eve could be a passive device, or Eve does not broadcast the MUI she is receiving. In addition, because MUI at Bob varies a lot (due to the behavior of other links) while the SIS factor does not, changing RxFJ according to MUI variations at Bob (in order to satisfy the rule in (15)) is not desirable. We show in the following that using a specific technique in setting TxFJ will help us to mitigate the dependence of the RxFJ (found from (15)) on MUI at Eve.

### B. PA Between TxFJ and Information Signals

After finding a suitable set of conditions/solutions for \( p'_q \) (i.e., the rule in (15)), we now focus on finding the optimal PA between TxFJ and information signals of Alice_q (i.e., \( \phi_q \)):

\[
\text{maximize } C_q^{\text{sec}} \\
\text{s.t. } 0 \leq \phi_q \leq 1.
\]

Note that although a simple one-dimensional search can find us the optimal value of \( \phi_q \), we would like to eventually solve (17) by avoiding the required knowledge of MUI at Eve. In this part of the paper, we propose a solution to solve (17) in the perfect E-CSI scenario. Later on, we show that our approach in the perfect E-CSI scenario is extendable to the case of unknown E-CSI, which has not been the case in recent designs.

Same as what we proposed for the RxFJ PA earlier, we approach problem (17) by first finding a bound on \( \phi_q \) that guarantees positive secrecy of link \( q \). Thus, the objective in (17) is assumed
to be positive, which reduces to
\[
\frac{\phi_q P_q}{a_q + b_q p_q'} > \frac{\phi_q P_q}{c_q + d_q p_q'}.
\] (18)
Simplifying this inequality, we end up with the following
\[
c_q > a_q + (b_q - d_q)p'_q.
\] (19)

The inequality in (19) is a bound on TxFJ power of Alice$_q$ (i.e., $\sigma_q$) because according to (9a), $c_q$ is a function of $\sigma_q$. Hence, reducing (19) gives us a bound for the portion of power allocated to the information signal (i.e., $\phi_q$) as well. However, for ease of presentation, we do not simplify (19) to write $\phi_q$ (or $\sigma_q$) at the left hand side of the inequality. We refer to (19) as the lower-bound on TxFJ power of link $q$ that provides it with positive secrecy.

To make use of the lower bound on TxFJ derived from (19), we first introduce a property of the secrecy rate of Alice$_q$

**Lemma 1.** If (19) is satisfied, the secrecy rate $C_q^{\text{sec}}$ is a monotonically increasing function of $P_q$ and $\phi_q$, respectively.

*Proof:* The inequality in (19) can be written as
\[
c_q = a_q + (b_q - d_q)p'_q + \delta
\] (20)
where $\delta > 0$ is a positive real value. Replacing the term $c_q$ in (10) with the right hand side (RHS) of (20), and taking the derivative of (10) (without the $\max\{\bullet\}$ operator) w.r.t. $P_q$ and $\phi_q$, we have
\[
\frac{dC_q^{\text{sec}}}{dP_q} = \frac{\phi_q \delta}{(a_q + \phi_q P_q + b_q p_q')(a_q + \phi_q P_q + b_q p_q' + \delta)}
\] (21a)
\[
\frac{dC_q^{\text{sec}}}{d\phi_q} = \frac{P_q \delta}{(a_q + \phi_q P_q + b_q p_q')(a_q + \phi_q P_q + b_q p_q' + \delta)}
\] (21b)
which are both positive and prove the lemma.

Recall that in setting RxFJ, which was done in (15), we observed that the optimal value of RxFJ ($p_q^*$) depends on MUI at Eve and Bob$_q$. In order to mitigate knowledge of MUI at Bob$_q$
and Eve in (15) (i.e., \(a_q\) and \(c_q\)), we examine the following alternative conditions for RxFJ:

\[
p'_q = \begin{cases} 
  P'_q, & \text{if } b_q < d_q \\
  0, & \text{if } b_q > d_q.
\end{cases}
\]  

(22)

Using the bound in (19), the following property shows the sufficiency of (22) to conclude (15).

**Proposition 1.** Provided that the following conditions hold, the RxFJ PA scheme in (22) is sufficient to conclude (15):

- \(c_q\) satisfies (19) and \(c_q > 0\).
- \((b_q - d_q)P'_q + \delta < 0\) when \(b_q < d_q\)

**Proof:** Assume that we use the conditions in (22) to decide on RxFJ power of link \(q\). Hence, we set \(p'_q = P'_q\) when \(b_q < d_q\). If \(c_q > 0\) and \(c_q\) satisfies (19) (first condition of Proposition 1), then \(c_q = a_q + (b_q - d_q)P'_q + \delta > 0\) when \(b_q < d_q\). Assuming that \((b_q - d_q)P'_q + \delta < 0\) (second condition of Proposition 1), one can conclude that \(a_q > c_q\), or equivalently \(a_q > a_q + (b_q - d_q)P'_q + \delta\). Hence, \(b_q < d_q\) is readily sufficient to deduce \(b_q < \frac{a_q d_q}{c_q}\) that appeared in (15). Similarly, \(b_q > d_q\) can be proven to be sufficient to satisfy \(b_q > \frac{a_q d_q}{c_q}\). Specifically, we set \(p'_q = 0\) according to (22). Hence, \(c_q\) must satisfy \(c_q = a_q + \delta\). Since \(\delta > 0\), \(a_q < c_q\). Therefore, \(b_q > d_q\) is sufficient to deduce \(b_q > \frac{a_q d_q}{c_q}\) that appeared in (15).

**Remark 1:** The key to conclude Proposition 1 are the assumptions \(c_q > 0\) and \((b_q - d_q)P'_q + \delta < 0\). Regarding the importance of \(c_q > 0\), if we have \(b_q < d_q\) and \(c_q = a_q + (b_q - d_q)P'_q > 0\), then \(b_q < d_q\) is sufficient to satisfy \(b_q < \frac{a_q d_q}{c_q}\), so both RxFJ PA schemes in (15) and (22) impose \(p'_q = P'_q\). However, when \(b_q < d_q\) (suggesting \(P'_q = P'_q\) in (22)) but \(c_q = a_q + (b_q - d_q)P'_q < 0\), we have \(b_q > \frac{a_q d_q}{c_q}\) (suggesting \(P'_q = 0\) in (15)). Hence, we have conflicting decisions made by (15) and (22). Condition \((b_q - d_q)P'_q + \delta < 0\) sets an upper bound on \(\delta\), i.e., \(0 < \delta < (d_q - b_q)P'_q\) if \(b_q < d_q\). According to (6) and (9), the terms \(b_q\) and \(d_q\) are in fact functions of self-interference channel, Alice-Bob channel, Bob-Eve channel, and Alice-Eve channels. Hence, if Proposition 1 holds, Bob only has to check whether or not

\[
\tau_q < \frac{|d_q^\dagger H_{qq} G_q|^2 |r_q^\dagger G_q'|^2}{|d_q^\dagger H_{qq}' G_q'|^2 |r_q^\dagger G_q|^2}
\]

(23)

\(^{16}\text{In the simulation section, we also observe the physical interpretation drawn from these conditions.}\)
to decide on RxFJ. In other words, (22) is sufficient to set the RxFJ power of Bob q. The intuitive interpretation from (23) is that the SIS factor needs to be small if the self-interference channel (i.e., |d_q d_{qq}|) has a large value, but if the Bob-Eve channel (i.e., |r_q r_{qq}|^2) is large enough, it can cancel out the effect of self-interference channel. In other words, Bob q must not use RxFJ if the self interference is not removed well enough. However, if Eve suffers more from the generated RxFJ, then Bob q can use it. Compared to (15), condition (22) is more desirable, as it does not impose real-time tracking of MUI (at Eve) to Bob q, thus simplifying the design. Combining (19) and (22), we have

\[
\begin{cases}
  c_q > a_q + (b_q - d_q)P'_q, & \text{if } b_q < d_q \\
  c_q > a_q, & \text{if } b_q > d_q
\end{cases}
\] (24)

Since the inequalities in (24) are strict, we write the following:

\[
\begin{cases}
  c_q = a_q + (b_q - d_q)P'_q + \delta, & \text{if } b_q < d_q \\
  c_q = a_q + \delta, & \text{if } b_q > d_q
\end{cases}
\] (25)

Overall, by mathematical manipulations done in equations (18)–(25), we convert problem (17) to the following problem:

\[
\begin{align*}
\text{maximize} & \quad C_{q}^{sec} \\
\text{s.t.} & \quad c_q = a_q + (b_q - d_q)P'_q + \delta \\
& \quad c_q > 0 \\
& \quad 0 < \delta < (d_q - b_q)P'_q + J(1 - t_q) \\
& \quad 0 \leq \phi_q \leq 1
\end{align*}
\] (26)

where \( P'_q \) in the first constraint is set according to (22), \( J \) is a sufficiently large positive number, and

\[
t_q = \begin{cases} 
1 & \text{if } b_q < d_q \\
0 & \text{if } b_q > d_q
\end{cases}
\] (27)

The first constraint in (83) is a constraint on \( \phi_q \) that imposes the optimal solution to yield positive
In other words, this constraint replaces the more general constraint in (17), such that we can ignore the \( \max\{\bullet\} \) operator in \( C_{q}^{\text{sec}} = \max\{C_q - C_{eq}\} \). This constraint together with the second and third constraints in (83) ensure us that setting \( p'_q^* \) according to (22) is sufficient to satisfy the more general conditions in (15) (but without the need to know MUI at Bob\( q \) and Eve). Note that \( t_q \) is a binary variable that is not among decision variables of (83), and can be easily computed by knowing \( b_q \) and \( d_q \).

Because \( c_q \) is a function of \( \phi_q \), one can simplify the first constraint in (83) to find the value of \( \phi_q \) that yields positive secrecy to the objective of (83). However, we still need to determine the value of \( \delta \) to ensure that such value found for \( \phi_q \) is the optimal for problem (83). A simple one-dimensional search in the interval defined by the third constraint in (83) can provide us with the best value of \( \delta \) and subsequently the optimal value of \( \phi_q \). To avoid additional computation imposed by the one-dimensional search process, we propose the following heuristic technique to set a value for \( \delta \). On the one hand, we do not wish to choose \( \delta \) near its upper bound due to the fact that the increase in \( \delta \) also increases the lower bound on TxJF which subsequently decreases the amount of power allocated to the information signal. On the other hand, the choice of \( \delta \) close to zero is also not desirable, as in (21b) the growth rate of secrecy rate would be decreased. Hence, we choose \( \delta = \frac{1}{2} |d_q - b_q| P'_q \). In simulations, we show that this heuristic choice of \( \delta \) yields a relatively close performance to that of the optimal solution found by one-dimensional search (see Fig. 4). In the next section, we use the derived conditions to model a secure power control game.

IV. GAME FORMULATION

In this section, using the ideas of Section III, we propose a power control scheme based on non-cooperative games.

A. Existence and Uniqueness of Nash Equilibrium

The first constraint in (83) can be written in general form as

\[
\begin{cases}
  c_q \geq a_q + (b_q - d_q) P'_q + \delta, & \text{if } b_q < d_q \\
  c_q \geq a_q + \delta, & \text{if } b_q > d_q.
\end{cases}
\]  

\[
(28)
\]

\(^{17}\)Note that the term \( c_q \) is a function of \( \phi_q \) (see (9)). An equivalent expanded version of this constraint is given in equation (29) later in this paper. In (83), however, for the sake of simplicity, we present this constraint in a more compact form.
\[ \phi_q \leq \max \left\{ \min \left\{ 1 - \frac{1}{P_q} \sum_{r \neq q}^{Q} \left( (A_{q,r} - B_{q,r}) \phi_r + C_{q,r} \phi_r + D_{q,r} \right) - p_q' \delta, 1 \right\}, 0 \right\} \]

(29)

\[
A_{q,r} = \frac{N_q - 1}{N_r - 1} \frac{|r_q^i G_q|^2}{|d_q^i H_{jq}|^2} \left( (N_r - 1)|d_q^i H_{jq}|^2 - |d_q^i H_{jq}|^2 \right) 
\]

(30a)

\[
B_{q,r} = \frac{N_q - 1}{N_r - 1} \frac{(N_r - 1)|r_q^i G_r|^2 - |r_q^i G_r|^2}{|r_q^i G_{jq}|^2} 
\]

(30b)

\[
C_{q,r} = \frac{N_q - 1}{N_r - 1} \frac{|r_q^i G_q|^2|d_q^i H_{jq}|^2 - |d_q^i H_{jq}|^2|r_q^i G_{jr}|^2}{|r_q^i G_{jq}|^2|d_q^i H_{jq}|^2} 
\]

(30c)

\[
D_{q,r} = (N_q - 1) \frac{|r_q^i G_q|^2|d_q^i H_{jq}|^2 - |d_q^i H_{jq}|^2|r_q^i G_r|^2}{|r_q^i G_{jq}|^2|d_q^i H_{jq}|^2} 
\]

(30d)

\[
E_q = (N_q - 1) \frac{\tau_q |r_q^i G_q|^2|d_q^i H_{jq}|^2 - |d_q^i H_{jq}|^2|r_q^i G_q'|^2}{|r_q^i G_{jq}|^2|d_q^i H_{jq}|^2} 
\]

(30e)

\[
F_q = (N_q - 1) \frac{|r_q^i G_q|^2}{|r_q^i G_{jq}|^2}. 
\]

(30f)

Simplifying (28) and taking into account other constraints of (83), an upper bound on \( \phi_q \) can be written as (29) where \( \delta = \frac{1}{2} |d_q - b_q| P_q' \) and the newly introduced notations in (29) are given in (30) at the top of the next page. Hence, link q’s optimization problem in (83), where \( q \in Q \), can be written as

\[
\text{maximize } \ C_q^{\text{sec}} \\
\text{s.t. } (29) 
\]

(31)

Hence, problem (83) is now reformulated as (31). With every legitimate link following such a strategy, the resulting interaction between them can be modeled as non-cooperative game [27] where the players are the links, the strategy set of the qth player is the set of constraints in (31), and the utility of each player is his secrecy rate. The result of Lemma 1, upon achieving positive secrecy for link q (i.e., satisfying the first constraint in (83)), the secrecy rate becomes

\[ \text{maximize } \ C_q^{\text{sec}} \\
\text{s.t. } (29) 
\]

(31)

Hence, problem (83) is now reformulated as (31). With every legitimate link following such a strategy, the resulting interaction between them can be modeled as non-cooperative game [27] where the players are the links, the strategy set of the qth player is the set of constraints in (31), and the utility of each player is his secrecy rate. The result of Lemma 1, upon achieving positive secrecy for link q (i.e., satisfying the first constraint in (83)), the secrecy rate becomes

18 Recall that in the discussion after (27), we proposed a heuristic method to determine the value of \( \delta \). The efficiency of this heuristic method is examined in the simulation section (see Fig. 4).
a monotonically increasing function of $\phi_q$. Hence, the that the best-response of the $q$th link, $q \in Q$, is when $\phi_q$ meets its upper bound in (29) with equality. The Nash equilibrium is a point at which no player is willing to unilaterally change his strategy given the strategies of other players [27].

Before going forward with our analysis, we need to emphasize that so far, our derivations were based on the assumption that link $q$, $q \in Q$, knows the MUI at Eve. In fact, although we were able to make RxFJ PA independent of MUI at Eve (by using (22)), setting the PA between TxFJ and information signal of Alice$_q$ as in (29) still requires knowledge of MUI at Eve. Such knowledge is possible if, for example, Eve feeds back the interference she is receiving to link $q$. However, this is an unrealistic assumption. Another possibility is when Eve is an active node, so link $q$ can acquire $G_q$ and $G_{j_q}$ from Eve’s activity. Then, the legitimate links can coordinate with each other and exchange these acquired channel gains along with their RxFJ powers and their PAs between TxFJ and information signals, so that link $q$ can use the best response in (29). While exchanging power levels can be possible via suitable control signaling schemes\(^{19}\), knowledge of E-CSI (i.e., $G_q$ and $G_{j_q}$, $\forall q \in Q$) at each link is a restricting assumption if not impossible, especially when Eve is a passive device. Nevertheless, we continue to analyze the game under full E-CSI in this section of the paper. This way, we can proceed with our analysis more easily. It also serves us as a benchmark to our next proposed scheme which is robust to uncertainties in knowledge of E-CSI. In other words, the robust scheme uses the same basic ideas behind the analysis under full E-CSI. The robust design is introduced in the next section of the paper.

The first game-theoretic analysis that we perform is to examine whether the game characterized by the optimization (31) admits a NE. Using the NE existence conditions of concave games, an NE exists if the strategy set of each player is non-empty, compact, and convex, and the utility function of each player is a continuous and (quasi-)concave function of its action, i.e., $C^\text{sec}_q$ is concave w.r.t. $\phi_q$ [30]. Convexity of each player’s strategy set in our game is easy to prove, thus omitted for the sake of brevity. Replacing the term $c_q$ with $a_q + (b_q - d_q)P'_q$ (as the first constraint in (83) suggests) and plugging it in (10), we take the second derivative of (10) w.r.t. $\phi_q$, which

\(^{19}\)Examples of such signaling protocols can be found in [28], [29].
can be expressed as
\[
\frac{d^2 C^\text{sec}_q}{d\phi^2_q} = P_q^2 \left( \frac{1}{a_q + \delta + \phi_q P_q + b P'_q} - \frac{1}{a_q + \phi_q P_q + b P'_q} \right)
\]  
(32)

which is always negative indicating that $C^\text{sec}_q$ is concave w.r.t. $\phi_q$.

A necessary and sufficient condition for the uniqueness of NE is proven in the following theorem.

**Theorem 1.** The game defined in (31) for all $q \in Q$ for which the best response of each player is set according to (29) (inequality (29) holds with equality), has a unique NE iff the following condition is satisfied:

\[
\rho(A + B) < 1
\]  
(33)

where $\rho(\bullet)$ indicates the spectral radius of a matrix (i.e., largest absolute value of eigenvalues of a matrix), $A$ is a $Q$-by-$Q$ matrix whose entries are written as

\[
A = \begin{cases} 
-\frac{P_r}{P_q} A_{q,r}, & r \neq q, \forall (r, q) \in Q \\
0, & r = q 
\end{cases}
\]  
(34)

and $B$ is as follows:

\[
B = \begin{cases} 
\frac{P_r}{P_q} B_{q,r}, & r \neq q, \forall (r, q) \in Q. \\
0, & r = q 
\end{cases}
\]  
(35)

with $A_{q,r}$ and $B_{q,r}$ defined in (30).

**Proof:** The uniqueness of NE can be proven by leveraging the concept of fixed-point theorem. In fact, if the iterative computation of each player’s best-response (i.e., $\phi_q$ meeting its upper bound in (29) with equality for all $q$) has a fixed point, the convergence point is the NE of the game [31]. We first analyze the existence of a fixed point for the argument inside $\max\{\min\{\bullet, 1\}, 0\}$ in (29). Then, we extend the analysis to include $\max\{\min\{\bullet, 1\}, 0\}$. Concatenating the constraint in (29) for all $q$ (without considering $\max\{\min\{\bullet, 1\}, 0\}$), the following fixed-point problem in its $t$–th iteration can be established:

\[
\Phi^{(t+1)} = T(\Phi^{(t)}) = \mathbb{1} + (A + B)\Phi^{(t)} + f
\]  
(36)

where $\Phi = [\phi_1, \ldots, \phi_Q]^T$, $\mathbb{1}$ is a vector of appropriate size whose entries are all 1, and $f$ is
a vector constructed by concatenating other terms in (29) for all \( q \). The rest of the proof is presented in Appendix A.

Remark 2: Using the condition in (33), the convergence of Jacobi iterative algorithm in the sense of [31, Ch. 2, Proposition 6.8] is guaranteed. In fact, at every iteration, all players simultaneously update their actions. Later on, we prove the convergence of our secure power control game under totally asynchronous updates (in the sense of [31, Ch. 6])\(^{20}\).

B. Algorithm Design

We now design an algorithm to implement the proposed power control game. Let \( T_q, \forall q \in Q \), be the set of iteration numbers when the \( q \)th link updates its action. For example, \( T_q = \{1, 3, 5\} \) indicates that the \( q \)th links performs the update in (31) in first, third and fifth iterations. Furthermore, Let \( \Theta_q^{(n)} = \{\theta_1^{(n)}, \ldots, \theta_Q^{(n)}\} \) denote the set of most recent times that the interference coming from each link is measured at Bob\(_q\) in the \( n \)th iteration. Hence, \( \theta_r^{(n)} \) is the most recent iteration in which the interference from the \( r \)th link, \( r \neq q \) is captured/updated, and \( \theta_{r,q}^{(n)} \leq n - 1 \). Therefore, in the \( n \)th iteration, the \( q \)th link, \( q \in Q \), performs the update in (31) based on \( \Theta_q^{(n)} \) if \( n \in T_q \). Using these definitions, we can now present an asynchronous algorithm that implements our proposed game, which is shown in Algorithm 1. Other suitable termination criteria can be used instead of the maximum iteration number. Special cases of the asynchronous scheme include Jacobi (or simultaneous) and Gauss-Seidel (or sequential) scheme [31]. The Jacobi scheme can be described as follows (\( q \in Q \)):

\[
T_q = \{1, 2, \ldots, \text{it}_{\text{max}}\}
\]

\[
\Theta_q^{(n)} = \{n - 1, \ldots, n - 1\}
\]

\(^{20}\)In [1], we only proved the convergence under Jacobi scheme.
where \( it_{max} \) is the maximum iteration number. In other words, in the Jacobi scheme, all links simultaneously update their action at each iteration. The Gauss-Seidel scheme can be described as follows:

\[
T_q = \{q, q + Q, q + 2Q, ..., q + \left( \frac{it_{max}}{Q} - 1 \right)Q \}
\]

\[
\Theta_{j}^{(n)} = \begin{cases} 
    \{n - (j - 1), ..., n - 1\} & \text{if } j = 1, \ldots, q - 1 \\
    \{n, n - (Q - 1), ..., n - q\} & \text{if } j = q, \ldots, Q 
\end{cases}
\]

which means that in each iteration, only one link updates its action, while all other links use their previously chosen actions. The following theorem guarantees the feasibility of asynchronous implementation of our proposed game:

**Theorem 2.** Algorithm 1 converges to the unique NE of the proposed game if Theorem 1 holds.

**Proof:** See Appendix B.

Note that (29) was derived only to proceed with the game-theoretic analysis of the problem. A detailed procedure to find the optimal value of \( \phi_q \) in a node is as follows. At a given iteration of our algorithm, say the \( n \)th iteration, after setting the optimal value of RxFJ, in order to determine the optimal PA, Bob\(_q\) needs to first measure the interference at his receive chain, i.e., \( a_q^{(n-1)} + b_q^{(n-1)}p_q^{*} \) must be measured, where \( a_q^{(n-1)} \) and \( b_q^{(n-1)} \) indicate the values of \( a_q \) and \( b_q \) at the previous iteration. Assuming that full knowledge of E-CSI is available, Bob\(_q\) also knows the MUI at Eve in the previous iteration, i.e., \( c_q^{(n-1)} + d_q^{(n-1)}p_q^{*} \) is known\(^{21}\). Hence, Bob\(_q\) does the following: 1) He subtracts the term \( \frac{|r^j_{G_{ej}}| \sigma_q}{|r^j_{G_{eq}}|^2} \) from \( c_q^{(n-1)} \); 2) He adds the result of subtraction to \( d_q^{(n-1)}p_q^{*} \). Denote the result of this addition as \( g_q \); 3) He finds the optimal PA in the \( n \)th iteration, which can be described as:

\[
\phi_q^* = \max \left\{ \min \left\{ 1 - \frac{|r^j_{G_{eq}}|^2}{|r^j_{G_{ej}}|^2} (a_q^{(n-1)} + b_q^{(n-1)}p_q^{*} - g_q), 1 \right\}, 0 \right\}. \tag{37}
\]

It can be seen that setting the optimal PA involves simple addition, subtraction and division of scalar values. Moreover, there is no need to know all interference terms at Bob\(_q\) and Eve, but only the aggregate of these terms (i.e., \( a_q \) and \( c_q \)) need to be known. Knowing the noise floor

\(^{21}\)Notice that throughout the iterations of our algorithm, \( b_q^{(n-1)} = b_q^{(n)} \) and \( a_q^{(n-1)} = d_q^{(n)} \). However, the values of \( a_q \) and \( c_q \) can vary across iterations.
at Bob, can be helpful to measure the interference level. For example, in 802.11 systems, the noise level usually stays at $-90$ dBm [32].

C. Discussion on Sufficient Conditions for NE Uniqueness

Although (33) is a tight condition, evaluating it requires knowledge of the whole matrix $A + B$, which is not desirable for distributed implementation. We introduce a sufficient condition which can be evaluated in distributed fashion. It is shown in [31, Proposition A.20] that for any induced matrix norm\(^{22}\) $|| \cdot ||$ and any square matrix $M$ we have $\rho(M) \leq ||M||$. Using this property, we consider the induced norm $|| \cdot ||$ to be $|| \cdot ||_\infty$, which is the infinity norm. Hence, assuming that $M$ is a $Q$-by-$Q$ matrix, a sufficient condition for $\rho(M) < 1$ is whether $||M||_\infty < 1$. Using this property in our game, a sufficient condition for our game to have a unique NE is whether

$$||A + B||_\infty = \max_q \sum_{r=1}^Q \frac{P_r}{P_q} |A_{q,r} - B_{q,r}| < 1. \quad (38)$$

The physical intuition drawn from the condition in (38) is not straightforward. One way to interpret this condition is to decompose this condition as follows: The term $A_{q,r}$ in (38) is mostly related to the MUI at each Bob which should be low enough, i.e., $|d^H_q h_{q,q}|$, $\forall q \in Q$ in $A_{q,r}$ should be large enough to guarantee the uniqueness of NE (see (30)). A sufficient separation between the links can satisfy this condition. The term $B_{q,r}$ in (38) is related to E-CSI components (see (30)). At first, it may seem that this condition requires each link to be the dominant interferer at Eve w.r.t. other links (i.e., $|r^H_q g_{j,q}|$, $\forall q \in Q$ in $B_{q,r}$ should be large enough). However, this is physically not possible. Instead, a more reasonable interpretation is that the NE is unique if every link has equal contribution in interfering with Eve’s reception. This way, the second summation in (38) would be constant which only depends on the number of interfering links, and the analysis of uniqueness is done via only the first summation. Since the nodes are not guaranteed to be positioned in a way that facilitates equal contribution in interfering with Eve, the only possibility is when Eve is far from all links. More discussion on this interpretation is given in the simulations section (see Fig. 5).

It can be seen that the uniqueness condition depends on the location of Eve because both $A_{q,r}$ and $B_{q,r}$ depend on Eve’s channels. Other studies such as [6], [9], [11] have also confirmed

\(^{22}\) The induced norm of matrix $M$ is defined as $||M|| \triangleq \max_{||x||=1} ||Mx||$ where $x$ is a vector and both norms on the RHS are vector norms.
the dependency of the unique NE (of non-cooperative secure power control games) on Eve’s channels. Such a coupling is neither practical (because E-CSI must be known) nor favorable (because Eve plays a role in the stability of the game). In what follows, we aim to mitigate knowledge of E-CSI and set both the NE uniqueness (derived in Theorem 1) free of Eve’s role. None of the approaches in [6], [9]–[11] were shown to be extendable to the case of unknown E-CSI. However, we show that our approach can be simply extended to cover the case of unknown E-CSI.

V. ROBUST POWER ALLOCATION GAME

In this section, we incorporate the assumption of unknown E-CSI in our game.

A. Computing the Best Response Under E-CSI Uncertainties

As knowledge of E-CSI becomes unknown, each legitimate link needs to ensure that positive secrecy is still preserved. Recalling the inequalities in (28) and (29), positive secrecy happens when $c_q > a_q + (b_q - d_q)p'_q$ or equivalently

$$
(1 - \phi_q)P_q > \psi_q + \tau_q p'_q E_q
$$

(39)

where

$$
\psi_q \triangleq \sum_{r=1}^{Q} \{(A_{q,r} - B_{q,r}) \phi_r P_r + C_{q,r} P_r + D_{q,r} p'_r\}.
$$

Under unknown E-CSI, for a given probability value $\varepsilon$, the $q$th link needs to satisfy the following:

$$
\Pr\{(1 - \phi_q)P_q > \psi_q + \tau_q p'_q E_q\} \geq \varepsilon.
$$

(40)

Using (22) and the Bayes law of total probability, we have

$$
\Pr\{(1 - \phi_q)P_q > \psi_q + \tau_q p'_q E_q\} =
\Pr\{b_q < d_q\} \Pr\{(1 - \phi_q)P_q \leq \psi_q + \tau_q p'_q E_q\} +
\Pr\{b_q > d_q\} \left(1 - \Pr\{(1 - \phi_q)P_q \leq \psi_q\}\right).
$$

(41)
\[
\phi_q \leq \max \left\{ \min \left\{ 1 - \Pr\{b_q < d_q\} \frac{E[\psi_q + \tau_q P'_q E_q]}{(1 - \varepsilon) P_q} - \Pr\{b_q > d_q\} \frac{E[\psi_q]}{(1 - \varepsilon) P_q}, 1 \right\}, 0 \right\}. \tag{44} 
\]

We assume that \(\psi_q + \tau_q p'_q E_q\) is a non-negative number for both values of \(p'_q\), i.e., \(\Pr\{\psi_q + \tau_q p'_q E_q > 0\} = 1\), otherwise (40) is always satisfied when \(\psi_q + \tau_q p'_q E_q < 0\), and Alice can spend all of the transmit power on information signal\[^{23}\]. Using Markov inequality in (41), the following holds

\[
\begin{align*}
\Pr\{b_q < d_q\} &\left(1 - \Pr\{(1 - \phi_q)P_q < \psi_q + \tau_q P'_q E_q\}\right) + \\
\Pr\{b_q > d_q\} &\left(1 - \Pr\{\psi_q\}\right) > \\
\Pr\{b_q < d_q\} &\left(1 - \frac{E[\psi_q + \tau_q P'_q E_q]}{(1 - \phi_q) P_q}\right) + \\
\Pr\{b_q > d_q\} &\left(1 - \frac{E[\psi_q]}{(1 - \phi_q) P_q}\right). \tag{42}
\end{align*}
\]

Hence, (40) remains true as long as we have

\[
\begin{align*}
\Pr\{b_q < d_q\} &\left(1 - \frac{E[\psi_q + \tau_q P'_q E_q]}{(1 - \phi_q) P_q}\right) + \\
\Pr\{b_q > d_q\} &\left(1 - \frac{E[\psi_q]}{(1 - \phi_q) P_q}\right) \geq \varepsilon. \tag{43}
\end{align*}
\]

Simplifying this inequality, we end up with (44) shown at the top of the next page. For the rest of this section, we explain how different terms in (44) can be computed. We first focus on computing \(\Pr\{b_q < d_q\}\). Using (6) and (9), we simplify \(b_q < d_q\), which is as follows

\[
b_q < d_q \Rightarrow |r_q^i G_q|^2 < \frac{|d_q^i H_{qq}|^2}{\tau_q |d_q^i H'_{qq}|^2} |r_q^i G'_q|^2. \tag{45}\]

The probability \(\Pr\{b_q < d_q\}\) can be written as

\[
\Pr\left\{ \frac{|r_q^i G_q|^2}{|r_q^i G'_q|^2} < \frac{|d_q^i H_{qq}|^2}{\tau_q |d_q^i H'_{qq}|^2} \right\}. \tag{46}\]

The small-scale fading components of \(r_q^i G'_q\) and \(r_q^i G_q\) are ZMCSRGVs with unit variances. Hence \(|r_q^i G_q|^2\) and \(|r_q^i G'_q|^2\) both have chi-square distributions with 2 and \(2N_q\) degrees of freedom,

\[^{23}\text{Intuitively, if Eve is not close no power needs to be allocated to TxFJ, hence suggesting that }\psi_q + \tau_q p'_q E_q < 0.\]
respectively. To tackle the issue of unknown large-scale fading components of $\mathbf{r}_q^\dagger \mathbf{G}_q'$ and $\mathbf{r}_q^\dagger \mathbf{G}_q$, we use stochastic geometry [33]. One can model nodes’ positions according to a spatial distribution, e.g., a Poisson point process (PPP). For instance, stochastic geometry has been used in modeling eavesdroppers’ positions in several recent works [34], [35]. We model the location(s) of Eve(s) according to an independent homogenous PPP, namely $\Omega$, with density $\lambda$. Such a representation can be used to model single or multiple Eves depending on the choice of $\lambda$.

Let $\Gamma_\gamma \triangleq |\mathbf{r}_q^\dagger \mathbf{G}_q|^2 - |\mathbf{r}_q^\dagger \mathbf{G}_q'|^2$ where $\Gamma$ and $\gamma$ are RVs that represent large-scale and small-scale fading components of $|\mathbf{r}_q^\dagger \mathbf{G}_q|^2 - |\mathbf{r}_q^\dagger \mathbf{G}_q'|^2$, respectively. Furthermore, let $\nu \triangleq \frac{|\mathbf{d}_q^\dagger \mathbf{H}_qq|^2}{\tau_q |\mathbf{d}_q^\dagger \mathbf{H}_qq'|^2}$. Using stochastic geometry and F-distribution, we have the following theorem:

**Theorem 3.** An analytical solution for (46) that is used in (44) is as follows:

$$
Pr\{\Gamma_\gamma < \nu\} = \exp\left(-\lambda \int_0^{d_0} \int_0^{2\pi} Pr\{\mathbf{s}_q^\gamma > \nu\} \beta \, d\beta d\varphi\right)
$$

where $\mathbf{s}_q \triangleq \left(\frac{\beta}{\sqrt{d_0^2 + \beta^2 - 2d_0\beta\cos\varphi}}\right)^\nu$ and $Pr\{\mathbf{s}_q^\gamma > \nu\} = (1 + \frac{\nu}{\mathbf{s}_q})^{-N_q}$.

**Proof:** See Appendix C.

We now turn our attention to $\mathbb{E}[\psi_q + \tau_q P_q^\dagger E_q]$ and $\mathbb{E}[\psi_q]$ in (44). We propagate the expectation in $\mathbb{E}[\psi_q + \tau_q P_q^\dagger E_q]$ to each term inside $\psi_q$ using (30). Hence, the expectation of the terms in (30) can be written as (48) shown at the top of the next page. Because the expectation terms in (48) contain non-negative RVs we can use the following identity:

$$
\mathbb{E} \left[ \frac{|\mathbf{r}_q^\dagger \mathbf{G}_q|^2}{|\mathbf{r}_q^\dagger \mathbf{G}_q'|^2} \right] = \int_0^\infty Pr\{\Gamma_\gamma > \nu\} d\nu
$$

where $Pr\{\Gamma_\gamma > \nu\}$ can be derived from Theorem 3. Other expectation terms that include E-CSI components can be treated the same as how we treat $\mathbb{E} \left[ \frac{|\mathbf{r}_q^\dagger \mathbf{G}_q|^2}{|\mathbf{r}_q^\dagger \mathbf{G}_q'|^2} \right]$.

While in the simulation section, we focus on the case where no knowledge on E-CSI components is available to links (i.e., both large-scale and small-scale fading parts of E-CSI components are not known), we can extract more insights from the derivations for unknown E-CSI by

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24 For example, if Eve is known to be distributed inside a certain region, we can find a suitable $\lambda$ (that represents the density as $\lambda$ Eves per unit of the surface area) such that the PPP matches our settings.

25 In [1], we assumed that the large-scale fading component of eavesdropper’s channels were known. However, in Theorem 3, we provided an analytical approach to cover the case of unknown large-scale fading components of E-CSI in our power control game.
considering the case where large-scale fading part of E-CSI is available. Hence, we can give a
close-form representation to (44). Knowledge of large-scale fading of Alice-Eve and Bob-Eve
channel is not new and has been assumed to be known for various scenarios. One example is
when Eve is acting as a reactive jammer. That is to say after some eavesdropping on the current
transmissions, Eve injects her jamming signal to disrupt the ongoing communications. In such a
case when jamming happens, assuming that the jamming power of Eve and the statistical features
of the jamming signal are previously known (e.g., PDF, mean), the legitimate links can measure
the jamming signal strength when it interferes with their transmissions. Hence, the approximate
location of Eve can be estimated. Moreover, in [36], it was shown that in a massive MIMO
scenario, a passive Eve might not be very dangerous and must therefore be active and attack
the training phase. This active attack can make Eve exposed, and hence the legitimate links can
acquire some knowledge about her location. Recently, the authors in [37] proposed a method
with which the legitimate nodes can detect the passive eavesdropper from the local oscillator
power leaked from its RF front end. Hence, an approximation on the location of Eve can be
acquired. Furthermore, the knowledge of large-scale fading was recently analyzed in [38] where
the directional properties (i.e., small-scale fading) of Eve(s) are unknown to Alice.

Regarding the calculation of \( \Pr\{b_q < d_q\} \) in (44), the small-scale part of \( X \triangleq \frac{|r_q^\dagger G_q|^2}{|r_q^\dagger G_q'|^2} \) in (46)
is equivalent to the SINR of a one-branch diversity combiner with \( N_q \) interferers [39, eq. (19)].
Thus,

\[ F_X(\xi) = 1 - \frac{1}{(1 + \xi)^{N_q}}. \] (50)

Using (50) in (46), we end up with

\[ \Pr\{b_q < d_q\} = 1 - \left(1 + \left(\frac{d_{qe}}{d_{qe}}\right)^\eta \frac{|d_q^t H_{qq}|^2}{|d_q^t H_{jjq}|^2}\right)^{-N_q}. \] (51)

To compute \( E[\psi_q + \tau_q P_q^T E_q] \) and \( E[\psi_q] \) in (44), we know that the small-scale fading part of random variables \(|r_q^t G_{jq}|^2\), \(|r_q^t G_{jr}|^2\), \(|r_q^t G_{r}|^2\), and \(|r_q^t G_{r}'|\)^2 have chi-square distributions with \(2(N_q - 1)\), \(2(N_r - 1)\), 2, and \(N_r\) degrees of freedom, respectively [13, Lemma 2]. Note that all of the aforementioned RVs are independent from each other because the precoding matrices \(V_q^{(1)}\) and \(V_q^{(2)}\), \(\forall q\) are unitary and orthogonal to each other (see Section II). The division of a (central) chi-square random variable by another independent (central) chi-square random variable has \(F\)-distribution [40]. Hence,

\[ E[A_{q,r}] = \frac{N_q - 1}{(N_r - 1)(N_q - 3)} \frac{(N_r - 1)|d_q^t H_{r}\|^2 - |d_q^t H_{jrq}|^2}{|d_q^t H_{qq}|^2} \] (52a)

\[ E[B_{q,r}] = 0 \] (52b)

\[ E[C_{q,r}] = \frac{N_q - 1}{(N_r - 1)(N_q - 3)} \frac{|d_q^t H_{jrq}|^2}{|d_q^t H_{qq}|^2} - \frac{N_q - 1}{N_q - 3} \frac{d_{re}}{d_{qe}} (-\eta) \] (52c)

\[ E[D_{q,r}] = \frac{N_q - 1}{N_q - 3} \left(\frac{|d_q^t H_{r}'|}{|d_q^t H_{qq}|^2} - \left(\frac{d_{re}}{d_{qe}}\right)^{(-\eta)}\right) \] (52d)

\[ E[E_q] = \frac{N_q - 1}{N_q - 3} \left(\frac{\tau_q |d_q^t H_{r}'|}{|d_q^t H_{qq}|^2} - \left(\frac{d_{re}}{d_{qe}}\right)^{(-\eta)}\right). \] (52e)

\[ E[F_q] = \frac{N_q - 1}{N_q - 3}. \] (52f)

The last issue is related to the on-off scheme proposed earlier for the RxJF. As it was shown in Section III, the choice of RxJF depends on whether \(b_q < d_q\) or \(b_q > d_q\). When \(b_q < d_q\) becomes a random variable in the case of unknown E-CSI, we choose to use RxJF whenever \(\Pr\{b_q < d_q\} > 0.5\).

\[\text{\footnotesize{Note that it is assumed that the knowledge of Alice-Bob channel, self-interference, and multi-user interference still hold.}}\]
B. Distributed Power Control Under E-CSI Uncertainties

With the derivations in (44)-(49), we can construct a game with the same structure as in Section IV where each link’s best response is computed from (44). Using the same logic behind Theorem 1, the following must hold to ensure a unique NE for the robust game:

$$\rho \left( \frac{E[A + B]}{1 - \varepsilon} \right) < 1$$  \hspace{1cm} (53)

where the expected value is element-wise. Note that $E[B_{q,r}] = 0$, so one can see that the analysis of the matrix $E[A + B]$ (see Section IV and Theorem 1), is simplified to $E[A]$. Therefore, the E-CSI is no longer present in NE uniqueness conditions. Moreover, for the $q$th link, $q \in Q$ to perform the PA scheme in (44), it requires the PA’s set by other links (i.e., $\phi_r$, $\forall r \in Q$, $r \neq q$), as well as the interfering channels between other legitimate links and Bob$_q$ (i.e., $H_{rq}$ and $H_{jrq}$, $\forall r \in Q$, $r \neq q$). Hence, no knowledge of MUI at Eve or E-CSI components is needed. Same as the previous section, an alternative condition to (53) is to replace the spectral radius with the infinity norm according to (38). Interestingly, the alternative condition for the robust game has a nice interpretation. Specifically, (53) is deduced if

$$\| \frac{E[A]}{1 - \varepsilon} \|_{\infty} = \max_{q} \sum_{r=1}^{Q} \frac{1}{1 - \varepsilon} |E[A_{q,r}]| < 1.$$  \hspace{1cm} (54)

Intuitively, if the interfering channels are small enough, a unique NE exists. Thus, the uniqueness conditions in the robust schemes are not dependent on E-CSI. The following asynchronous algorithm implements the robust version of our game:

**Algorithm 2** Asynchronous Iterative Secure Power Allocation (robust version)

1: Given $\varepsilon$, calculate (46) and set $p'_q = P'_q$ if $\Pr\{b_q < c_q\} \geq 0.5$, or $p'_q = 0$ if $\Pr\{b_q < c_q\} < 0.5$.
2: for $n=1$ to maximum iteration do
3: \hspace{1em} Set $\phi_q^{(n)} = \begin{cases} \text{Equal to RHS of (44)}, & \text{if } n \in T_q, \forall (q) \in Q, \\ \phi_q^{(n-1)}, & \text{otherwise} \end{cases}$
4: end for

Note that the case where large-scale fading part of E-CSI is known was only mentioned to give intuitions on the behavior of the robust scheme. In simulations, we only consider the case where neither the large-scale nor the small-scale fading parts of E-CSI are known.
VI. NUMERICAL RESULTS

In this section, we verify our theoretical analyses. We show our results for a four-link network\(^{28}\). Eve is located at \((X_e, Y_e)\) on a 2-D coordinate system. Alices are randomly placed on the boundary of a circle, known as simulation region, with radius \(r_{\text{circ}}\) whose center is at the origin of the coordinate system. Specifically, only the phase of Alices’ placements has uniform distribution. Each Alice has a fixed distance (communication range) with her corresponding Bob denoted as \(d_{\text{link}}\)! Each Bob is placed randomly around his corresponding Alice on the boundary of a circle whose center is the location of Bob’s corresponding Alice with radius \(d_{\text{link}}\). Again, only the phase of Bobs’ placements has uniform distribution. The noise level is set to 0 dBm. Unless stated otherwise, the power constraint for each legitimate link is set to \(P_q = 20\) dBm, \(\forall q\), the maximum RxFJ power at each Bob is \(P'_q = 15\) dBm, \(\eta = 2.5, \tau_q = -100\) dB\(^{30}\), \(d_{\text{link}} = 10\) m, and finally Jacobi algorithm is used in all simulations. Regarding the unknown location for Eve, Bob\(_q\) assumes that Eve is distributed in a circle around him with radius \(r_0 = 5\) m according to a PPP with \(\lambda = \frac{1}{25\pi}\) Eve/m\(^2\), \(q \in Q\).

For the first numerical result, We set up our system model in the presence of an eavesdropper where the PA between TxFJ and information signal for all links is set to \(\phi = 0.5\). We aim to find out if the RxFJ PA scheme in (15) is sufficiently close to an optimal scheme to solve (12). To do so, we perform the optimal assignment of RxFJ power for (12) with a simple one-dimensional search method for several channel realizations and count the times when the solution found from one-dimensional search reduces to the solution in (15). In Fig. 2, we plot the probability of having both positive secrecy and the optimal value of RxFJ power for problem (12) (found from a one-dimensional search) being either the maximum or zero according to the scheme in (15) for all links. Such probability shows how frequent the scheme in (15) gives us the optimal value of RxFJ power. It can be seen in Fig. 2 that this probability is very high even for the cases where the power budget for RxFJ is high. Also, it can be seen that the size of simulation region has a negligible effect on this probability.

Fig. 3 (a)-(c) show the number of links that use RxFJ in the network for the two RxFJ PA schemes derived in (15) and (22) where in (15), \(c_q\) is set according to (25). We assumed that

\(^{28}\)The results for this case can be generalized to arbitrarily larger number of links.

\(^{29}\)Using a common communication range is a generic assumption in wireless ad hoc networks [34], [35].

\(^{30}\)Such an SIS factor was reported in recent practical implementations of full-duplex radios [17].
all links use \( \phi_q = 0.5 \) as the PA for information and TxFJ signals. It can be seen from these figures that using the RxFJ PA in (22) has a close performance to (15) whenever Alices’ power budgets are high enough (see Fig. 3(a)) or when Bobs’ RxFJ power budgets are low enough (see Fig. 3(b)). Examining \( c_q \) in (25), one can easily see that low transmit powers would decrease \( a_q \) and high RxFJ powers would increase \( (b_q - d_q)P_q' \). Both of these situations are detrimental to the scheme in (22), as they violate the condition \( c_q > 0 \) which is a requirement for sufficiency of the scheme in (22) (See Proposition 1 and Remark 1). Using high enough power budgets at Alices (i.e., \( P_q, \forall q \in Q \)) and low enough RxFJ powers at Bobs (\( P_q', \forall q \in Q \)) for all links can ensure that \( c_q \) will remain positive. As it can be seen in Fig. 3 (c), for a suitable choice of transmit power and RxFJ power, both conditions stay close to each other regardless of number of links in the network. Overall, under high enough transmit power budget and low enough RxFJ power budget, the sufficient condition (22) yield a performance equivalent to (15) and sets Alice free of having to track the MUI at Bob and Eve.

Next, we compare the performance of our proposed methods for PA between TxFJ and information signals. Specifically, in one method, we use one-dimensional search to find the
best value of $\delta$ in (83). In the other method, we use our proposed heuristic method for finding $\delta$, i.e., $\delta = \frac{1}{2}|d_q - b_q|P'_q$. We compare the resulting secrecy sum-rate of these two methods in Fig. 4. It can be seen that the proposed heuristic method has a very close performance to that of the one-dimensional search, suggesting that we can use the heuristic method for assigning $\delta$ without imposing the relatively larger computational complexity of the one-dimensional search method.

Fig. 5 shows the probability of satisfying the uniqueness conditions derived in (33) and (38) for a two-link scenario with full knowledge of E-CSI. The vertical axis at the left of each subfigure indicates the probability of satisfying (33), i.e., $\rho(A + B) < 1$. Specifically, each point on the curve related to (33) (indicated by $n_1$) is the result of averaging the number of times (33) holds over 100 network topologies where in each topology 500 channel realizations are simulated and averaged. Thus, the probability of convergence for (33) is $n_1/(100 \times 500)$ where $n_1$ denotes the number of times that (33) is satisfied over all network topologies and channel realizations. Let $n_2$ denote the number of times that condition (38) is satisfied given that (33) is already satisfied. Hence, the vertical axis at the right of each subfigure indicates the ratio $n_2/n_1$ for which we have $n_2/n_1 < 1$, since $n_2$ counts the times (38) is true among the times (33) holds.

The horizontal axis in Fig. 5 indicates the value of $X_e$. While the value of $Y_e$ is fixed for a subfigure, it is different from one subfigure to another. For the two-link case, condition (33) is highly probable in all scenarios. The practical condition in (38), however, is only good when Eve is relatively far from the network, but as Eve becomes closer to the network this condition is less efficient. Interestingly, as Eve approaches the origin, for $Y_e = 0$ in Fig. 5 (a), the probability of satisfying (38) increases. The reason for such a result is because of the simulation model, which

\cite{31}Note that the one-dimensional search is in fact the optimal approach in solving (83).
verifies the physical interpretation given for (38). In fact, the origin is where the distance of all
links to Eve is the same because the simulation model puts all of Alices in the boundary of the
simulation region which is a circle. One can see that when the y-coordinate of Eve changes in
Fig. 5 (b) and Fig. 5 (c), the location $X_e = 0$ becomes more similar to other points inside the
simulation region. We did not however, see this phenomenon for higher number of links, which
is attributed to the fact that the second summation in (38) becomes too large with high number
of links, even though it is a constant for when $(X_e, Y_e) = (0, 0)$.

Fig. 6 (a) shows the variation of convergence probabilities of robust and full E-CSI methods
w.r.t. $r_{circ}$ for the four-link case. The convergence probability is calculated as number of times
the conditions in (33) and (38) (indicated by “full E-CSI, $n_1$” and “full E-CSI, $n_2$”, respectively,
in Fig. 6 (a)), and their equivalents for the robust game (i.e., (53) indicated by “Robust, $n_1$” and
(54) indicated by “Robust, $n_2$”) hold true divided by the number of channel realizations. It can
be seen that for the case of full E-CSI, probability of uniqueness of NE using (38) is very low.
However, in the case of unknown E-CSI, since the nodes are indifferent w.r.t. E-CSI, far less
restrictive conditions than that of full E-CSI scenario can be achieved. In fact, although the
distances between links and Eve become larger as $r_{circ}$ grows, the uniqueness of NE in the
full E-CSI case still remains unpredictable. Specifically, in (29) it can be seen that the distances
between legitimate links and Eve exist in both nominator and denominators of the terms described
in (30). This is the main reason that by increasing $r_{circ}$, the convergence of our algorithm in the
full E-CSI case is not significantly affected. On the contrary, in the robust method, by increasing
the radius of simulation region, interference at each Bob becomes weaker. So, as the physical
interpretation mentioned for (54) suggested, the NE uniqueness becomes more often. Moreover,
in robust version, as $\varepsilon$ becomes larger, the uniqueness conditions become more restrictive, which
is in line with the derivation in (53).

Fig. 6 (b) shows the convergence probability versus number of transmit antennas at each
link. It can be seen that the probability of NE uniqueness becomes higher with more transmit
antennas. The reason is that as the number of antennas grows, the communication channels
become almost deterministic and only dependent on large-scale fading components. Specifically,
looking at (29), (30), (44), it can be shown that for large number of antennas, the values of
$|r_q^\dagger G_q|^2$, $|d_q^\dagger H_{qq}|^2$, and $|d_q^\dagger H_{rq}|^2$ would be close to their large-scale fading components multiplied
by 2, as their distributions are chi-squares with two degrees of freedom [13, Lemma 3]. Using
the same argument, the values of $|r_q^\dagger G_{qq}|^2$, $|d_q^\dagger H_{rq}|^2$, and $|r_q^\dagger G_{rr}|^2$ will be asymptotically close
Fig. 5: Probability of convergence vs. eavesdropper’s location for the full E-CSI case: (a) \( Y_e = 0 \), (b) \( Y_e = 10 \) m, (c) \( Y_e = 40 \) m, \( r_{c,ref} = 30 \) m, \( N_q = 8, M_q = L = 1, \forall q, Q = 2 \).

to their large-scale fading component multiplied by certain factors, as \( |r_q^rG_{rqr}|^2 \), \( |d_q^H_{hrq}|^2 \), and \( |r_q^rG_{rqr}|^2 \) have chi-square distributions with \( 2(N_q - 1) \), \( 2(N_r - 1) \) and \( 2(N_r - 1) \) degrees of freedom, respectively. Hence, the values of \( A_{q,r} \) and \( B_{q,r} \) are both close to zero which result in small eigenvalues and thus satisfying the convergence condition in (33) more often. The same argument holds for conditions in (38), (54), i.e., they become more probable as the number of transmit antennas grow.

Fig. 7 (a)-(c) show the achieved secrecy sum-rate of robust and full E-CSI methods as well as the globally optimal solutions of the secrecy sum-rate maximization vs. the radius of our simulation region. Furthermore, Fig. 7 (d)-(f) show the resulting sum of information and leaked rates of our methods vs. the radius of our simulation region. The maximum amount of iterations for Algorithm 1 and 2 are set to 50 iterations. We conducted this simulation for two scenarios: 1) when Eve uses MRC decoder, and 2) when Eve uses MMSE decoder.

We also have two baseline schemes in Fig. 7(a)-(c). First, the scheme where no RxFJ is used at Bob, and second, the scheme where no TxFJ is used at Alices. From Fig. 7 (a)-(c), it can be seen that our approaches have less secrecy compared to globally optimal solutions. The reason is that the NEs of our proposed game are not necessarily guaranteed to be globally optimum for the secrecy sum-rate. NE is a stable point where no link can gain more secrecy given the interference received from other nodes. In other words, the NE of our game is unilaterally optimum for each link. Furthermore, both cases of the robust method have less secrecy sum-rates than that of the full E-CSI method.

Furthermore, it can be seen that both no RxFJ and no TxFJ schemes have significantly less

\[32\] Although our analysis was limited to the case of using MRC decoder at Eve, we still observed the convergence of our algorithm for the case of MMSE decoder.
Fig. 6: (a) Probability of convergence vs. $r_{\text{circ}}$ ($X_e = Y_e = 5$, $N_q = 8$, $M_q = L = 5$, $\forall q, Q = 4$), (b) Probability of convergence vs. $N_q$ ($X_e = Y_e = 0$, $r_{\text{circ}} = 15$ m, $M_q = L = 5$, $\forall q \in Q$)

secrecy sum-rate compared to our approaches, which signifies the importance of FJ. Lastly, in our model, it seems that using no TxFJ affects the secrecy sum-rate more than using no RxFJ. The performance of both of these gets worse when Eve employs MMSE receiver, which is not shown here due to space limitations.

Although our analysis was limited to the case of using MRC decoder at Eve (see Section II), we still observed the convergence of our algorithm for the case of MMSE decoder. One reason that we did not analyze the case of MMSE receivers at legitimate links or Eve is that MMSE receivers add to the complexity of links’ best responses. In fact, in addition to the TxFJ and RxFJ powers being updated at each iteration of the game, the MMSE receiver needs to be updated at each iteration of the game as well, thus increasing the complexity of a link’s actions. In contrast, using the MRC decoder employed at Eve/Bobs allows us to only focus on TxFJ and RxFJ PA.

Another reason for choosing MRC over MMSE receivers is the amount of computational complexity that MMSE imposes on legitimate nodes which can be significant depending on the number of Alice’s/Bob’s antennas. The first step to compute an MMSE receiver at each Bob is to calculate the covariance matrix of the interference at the receive chain of each Bob, which is basically a vector multiplication operation. Then, each Bob needs to measure the channel between himself and his corresponding Alice and multiply it to the inverse of the interference’s covariance matrix to establish the MMSE receiver. Compared to the MRC receiver, which only requires the channel between Bob and his corresponding Alice, the MMSE receiver requires three more operations (two matrix multiplications and one inverse) to be established$^{33}$.

According to Fig. 7 (d)-(e), for a given $\varepsilon$ in the robust method, regardless of the decoder

$^{33}$We skipped the detailed description of an MMSE method of reception for the sake of brevity. The fundamentals of MMSE receivers can be found at [22, chapter 6]. More comparison between MRC and MMSE receivers is given in the rest of this section.
at Eve, the sum of information rates remains the same, which indicates that the interference management between legitimate links in the robust method is completely decoupled from Eve characteristics. In other words, in the robust method, the nodes are indifferent to E-CSI. Moreover, for when $\varepsilon = 0.9$, the leaked rate is significantly reduced compared to when $\varepsilon = 0.1$ because the probability of achieving positive secrecy is set to be higher for when $\varepsilon = 0.9$. However, the penalty for achieving positive secrecy with high probability (in the robust method) is that the nodes cannot manage interference between themselves as efficiently as in the full E-CSI case or the case where $\varepsilon = 0.1$. This leads to having lower sum of information rates when $\varepsilon = 0.9$. Another penalty of choosing $\varepsilon$ to be too high is that the uniqueness of NE becomes less often. As it was shown in Fig. 6 (a), by increasing $\varepsilon$, the probability of NE uniqueness in the robust scheme decreases. The issue of not having a unique NE can lead to instability.
of our power control algorithm, as this algorithm is not guaranteed to converge to any point if the NE uniqueness conditions are violated. Hence, the iterations of our algorithm must be manually terminated with no sign of convergence. It can also be seen in Fig. 7 (a)-(c) that when Eve employs MMSE decoder, the secrecy sum-rate decreases more than when MRC decoder is adopted. The reason is that the MMSE receiver takes into account the effect of interference, so Eve is able to mitigate a part of interference on herself. According to (29), in the full E-CSI case, the action of a link depends on the decoder that Eve uses. Therefore, we can see in Fig. 7 (f) that the sum of information rates for the full E-CSI case is affected when Eve uses MMSE decoder. As the radius of simulation region increases, the SINR at Eve decreases, so it is well-known that for the low SINR regime, the MMSE receiver must reduce to the MRC receiver [22]. This phenomenon can be seen in Fig. 7 (d)-(f) where the leaked rate at Eve is the same for both decoders for a large enough $r_{\text{circ}}$.

Fig. 7 (g)-(i) show that in all approaches secrecy sum-rate grows as $P_q$ increases. Hence, by using RxFJ and TxFJ, positive secrecy and arbitrary secrecy levels (by changing the links’ transmit powers) are achievable, thus extending the same property that existed in the single-user scenario [2]. We also verified such a scaling at the per-link level. Same as what was discussed above, the secrecy sum-rate achieved for the full E-CSI method is larger than that of the robust methods. Also when $\varepsilon$ is chosen to be too large, the nodes are not able to do an efficient interference management. This could sound as a counter-intuitive result. In fact, a low value of $\varepsilon$ increases the probability of having zero secrecy (see Section V). However, one can see that the performance of the robust approach is still acceptable for when $\varepsilon = 0.1$. We conjecture that this could be due to the F-distribution, which was the distribution involved in reducing the best responses in (29) to (44). The F-distribution is a positively skewed distribution. Hence, most of the density of the distribution of best response is concentrated at the left of the median, and the mean is most likely at the left of the median. This means that to ensure that 

$$(1 - \phi_q)^{\frac{1}{1 - \varepsilon}} P_q \left( \frac{b_q}{d_q} \right) \left( 1 - \frac{1}{P_q} \right) + \left( 1 - \frac{1}{1 - \varepsilon} \right) \Phi_q \right),$$

the density of the values above the mean is relatively low, suggesting that the PA factor in (44) will be most likely providing positive secrecy.

Fig. 8 shows the convergence of the proposed asynchronous algorithm under different update schemes for a sample settings where the NE is unique. All update schemes converge to the same point, indicating the uniqueness of NE. The Jacobi method converges faster due to simultaneous updates for all users at each iteration. Fig. 8 (c) also shows convergence under random updates.
for each link. Each link generates a random integer in interval $[2, 6]$ that specifies the number of iterations after the current iteration when the link updates his action (i.e., the update pattern denoted by $T_q$, see Section IV.B). Clearly, introducing asynchronism in the system degrades the convergence speed, but the system will eventually converge to the NE. It can also be seen from the figures of the bottom and above rows that the convergence occurs with and without $\min\{\max\{x, 0\}, 1\}$ operator. Clearly, such an operator does not impact the convergence.

VII. CONCLUSIONS

In this paper, we proposed a game-theoretic approach for power control in an interference network tapped by an external eavesdropper. We proposed a framework under which every link can utilize both RxFJ and TxFJ to achieve a positive secrecy rate. Next, we modeled the interaction between the players as a game and derived sufficient conditions for the uniqueness of the resulting NE. We also proposed an asynchronous algorithm that can implement the proposed game and is robust to practical issues in the network such as transmission delays. Next, we proposed another version of our game that is robust to when the eavesdropping channels are unknown. We showed in simulation that our proposed approach for achieving positive secrecy using TxFJ and RxFJ are efficient enough to be considered as best responses for legitimate links. Moreover, the performance of robust schemes are close to the one that assumes knowledge of E-CSI. Lastly, the secrecy sum-rate scales with the power budget at legitimate transmitters, regardless of the knowledge of E-CSI. Analytical characterization of the proposed game when Eve uses MMSE decoder could be an interesting subject of future research.
REFERENCES


APPENDIX A

PROOF OF THEOREM 1

Using [31, Proposition 6.1], the fixed point iteration in (36) converges to a point $\phi^*$ from any initial point iff $\rho(A + B) < 1$. We now introduce the following theorem
Theorem 4. [31, Ch. 2, Proposition 6.6] For any square matrix $M$ and any $\epsilon > 0$, there exists an induced norm, $\| \cdot \|$ such that $\rho(M) \leq \|M\| \leq \rho(M) + \epsilon$. \hfill $\Box$

Using the above theorem, since $\rho(A + B) < 1$, we can choose $\epsilon > 0$ arbitrarily close to zero such that $\rho(A + B) + \epsilon < 1$. Hence, we can find an induced norm $\|A + B\|$ such that $\|A + B\| \leq \rho(A + B) + \epsilon$. Therefore, we are able to convert the condition $\rho(A + B)$ to an equivalent condition based on an induced norm, i.e., $\|A + B\|$. We use this result later during this proof. To proceed with further analysis, we need the following definition:

Definition 1. [31] Consider the following iteration:

$$\Phi^{(t+1)} = T(\Phi^{(t)}), \quad t = 1, 2, ..., (55)$$

where $T$ is a mapping from $A$ (a subset of $\mathbb{R}^Q$) to itself, and $t$ indicates the index of iterations. If $T$ is continuous and

$$\|T(\Phi^{(1)}) - T(\Phi^{(2)})\| \leq \Omega \|\Phi^{(1)} - \Phi^{(2)}\|, \quad \forall \{\Phi^{(1)}, \Phi^{(2)}\} \in A^2, (56)$$

where $\|\cdot\|$ is a norm in $A$ and $\Omega \in [0, 1)$, then the mapping $T$ is a contraction mapping with $\Omega$ as the contraction modulus, and sequence $\{\phi^{(t)}\}$ generated by iterations in (55) converges to the fixed point $\phi^*$.\hfill $\Box$

Using this definition and the result of Theorem 4, we can show the iteration in (36) as a contraction mapping, i.e.,

$$\|T(\Phi^{(1)}) - T(\Phi^{(2)})\| \leq \|(A + B)(\Phi^{(1)} - \Phi^{(2)})\|$$

$$\leq \|A + B\| \|\Phi^{(1)} - \Phi^{(2)}\| \leq \rho(A + B) + \epsilon < 1 (58)$$

where $\|A + B\| < 1$, (58) is due to Cauchy-Schwartz inequality, and the induced norm $\| \cdot \|$ is chosen such that for some $\epsilon > 0$ we have $\|A + B\| \leq \rho(A + B) + \epsilon < 1$ (cf. Theorem 4). This result will be used later in this proof.

We now focus on $\min\{\cdot\}$ and $\max\{\cdot\}$ functions. The operator $\max\{\min\{\phi_0, 1\}, 0\}$, for some $\phi_0 > 0$, can be equivalently shown as a Euclidean projection. Specifically, the Euclidean

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34For the sake easy presentation, we omitted introducing the weighted norm, while this is the type of norm used in [31, Ch. 2, Proposition 6.6]. Nevertheless, all of our analyses can be extended to the case of weighted norms as well.
projection of a scalar $\phi_0$, denoted as $[\phi_0]^+$, can be written as the following optimization problem

$$\minimize_{\tilde{\phi}} \ ||\tilde{\phi} - \phi_0||^2$$

s.t. $0 \leq \tilde{\phi} \leq 1$.  \hfill (59)

The KKT conditions of this problem are written as follows:

$$\tilde{\phi} - \phi_0 - \nu + \lambda = 0,$$  \hfill (60)

$$\nu \geq 0, \tilde{\phi} \geq 0, \nu \phi = 0$$  \hfill (61)

$$\lambda \geq 0, \tilde{\phi} \leq 1, \lambda(\phi - 1) = 0;$$  \hfill (62)

If $\nu > 0$, then $\tilde{\phi} = 0$. Hence, $\lambda = 0$ and we have $\nu = -\phi_0$, or equivalently $\phi_0 \leq 0$. If $\lambda > 0$, then $\tilde{\phi} = 1$. Hence, $\nu = 0$, and we have $1 + \lambda = \phi_0$, or equivalently $\phi_0 \geq 1$. If $\lambda = 0$ and $\nu = 0$, then $0 \leq \tilde{\phi} \leq 1$. Hence, $\tilde{\phi} = \phi_0$. Summarizing these conditions, we have

$$\tilde{\phi}^* = \argmax_{0 \leq \tilde{\phi} \leq 1} ||\tilde{\phi} - \phi_0||^2 = \begin{cases} 0, & \text{if } \phi_0 \leq 0, \\ 1, & \text{if } \phi_0 \geq 1, \\ \phi_0, & \text{if } 0 \leq \phi_0 \leq 1. \end{cases}$$  \hfill (63)

The right hand side of (63) is exactly the definition of the operator $\max\{\min\{\bullet, 1\}, 0\}$.

Converting $\max\{\min\{\bullet, 1\}, 0\}$ to Euclidean projection, we use the non-expansive property of Euclidean projection which is as follows [31, Ch. 3, Proposition 3.2]:

$$\left\| \left[ \mathcal{T}(\Phi(1)) \right]^+ - \left[ \mathcal{T}(\Phi(2)) \right]^+ \right\| \leq \left\| \mathcal{T}(\Phi(1)) - \mathcal{T}(\Phi(2)) \right\|$$  \hfill (64)

The non-expansive property of Euclidean projectors can be generalized to all vector norms because all vector norms (i.e., norms in $\mathbb{R}^n$) are equivalent, i.e., for any two different norm $|| \bullet ||^1$ and $|| \bullet ||^2 \exists \alpha_1 \in \mathbb{R}$ and $\alpha_2 \in \mathbb{R}$ such that $\alpha_1 ||x||^1 \leq ||x||^2 \leq \alpha_2 ||x||^1, \forall x \in \mathbb{R}^n$ [41]. Hence, we have the following chain of inequalities

$$\left\| \left[ \mathcal{T}(\Phi(1)) \right]^+ - \left[ \mathcal{T}(\Phi(2)) \right]^+ \right\| \leq \left\| \mathcal{T}(\Phi(1)) - \mathcal{T}(\Phi(2)) \right\|$$  \hfill (65)

$$\leq \left\| (\mathbf{A} + \mathbf{B})(\Phi(1) - \Phi(2)) \right\| \leq ||\mathbf{A} + \mathbf{B}|| ||\Phi(1) - \Phi(2)||$$  \hfill (66)
Hence,
\[ \left\| \left[ \mathcal{T}(\Phi(1)) \right]^+ - \left[ \mathcal{T}(\Phi(2)) \right]^+ \right\| \leq \| A + B \| \left\| \Phi(1) - \Phi(2) \right\|. \tag{67} \]

Setting the norm in (67) as the same norm in (58), the best response of each player is a contraction map, and thus has a unique fixed point (NE).

**APPENDIX B**

**PROOF OF THEOREM 2**

Similar to the proof of Theorem 1, consider the following iteration:

\[ \Phi^{(t+1)} = \mathcal{T}(\Phi^{(t)}), \ t = 1, 2, \ldots. \tag{68} \]

We use the asynchronous convergence theorem [31], which is as follows:

**Theorem 5.** The iteration in (68) converges asynchronously if the following conditions are satisfied:

1) There exists a sequence of nonempty sets \( \mathcal{X}(t) \) such that
\[ \cdots \subset \mathcal{X}(t+1) \subset \mathcal{X}(t) \subset \cdots \subset \mathcal{X}. \tag{69} \]

2) The iteration \( \mathcal{T}(\cdot) \) must satisfy \( \mathcal{T}(\Phi^{(t)}) \in (t+1) \). Furthermore, every limit point of \( \Phi^{(t)} \) must be a fixed point of \( \mathcal{T}(\cdot) \).

3) For every \( t \), we must have \( \mathcal{X}(t) = \mathcal{X}_1(t) \times \cdots \times \mathcal{X}_Q(t) \) where \( \mathcal{X}_q(t) \subset \mathcal{X}_q, \ q \in Q \).

\[ \Box \]

The first item of Theorem 5 can be proven as follows. let \( \Phi^* = [\Phi^*_1, \ldots, \Phi^*_Q]^T \) be the fixed point of the iteration in (68). Consider the following set

\[ \mathcal{X}_q(t) = \{ \Phi \in \mathbb{A} : \| \Phi - \Phi^* \|_{2, \text{block}} \leq \alpha^t \| \Phi^{(0)} - \Phi^* \|_{2, \text{block}} \} \subset \mathbb{A} \tag{70} \]

where \( \mathbb{A} = \{ \Phi \in \mathbb{R}^Q : 0 \leq \Phi \leq 1 \} \), \( \| a \|_{2, \text{block}} = \max_{q \in Q} |a_q|_2 \) is the vector block-maximum norm for \( a = [a_1, \ldots, a_Q]^T \) with \( \| \cdot \|_2 \) defined as the Euclidean norm, and \( \alpha = \| A + B \| \) with \( A \) and \( B \) defined in (34) and (35). It can be easily seen that iff \( \alpha < 1 \) we have

\[ \alpha^{t+1} \| \Phi^{(0)} - \Phi^* \|_{2, \text{block}} < \alpha^t \| \Phi^{(0)} - \Phi^* \|_{2, \text{block}}, \ \forall n = 0, 1, \ldots. \tag{71} \]
Hence, we can conclude that

\[ X(t+1) \subset X(t) \subset \Lambda, \quad t = 1, 2, \ldots \]  

(72)

The second item of Theorem 5 can be concluded from Theorem 1. As for the third item of Theorem 5, consider the following. The set \( X(t) = X_1(t) \times \cdots \times X_Q(t) \) can be decomposed as follows for all \( t \):

\[ X_q(t) = \{ 0 \leq \Phi_q \leq 1 : ||\Phi_q - \Phi_q^*|| \leq \alpha^t ||\Phi^{(0)}(0) - \Phi^*||_{2, \text{block}} \} . \]  

(73)

Hence, all three conditions required for asynchronous convergence of Algorithm 1 can be satisfied provided that Theorem 1 holds.

APPENDIX C

PROOF OF THEOREM 3

Without loss of generality assume that \( \Omega \) represents a set of multiple independent (fictitious) Eves, whose locations (inside a given area) follows the PPP distribution with density \( \lambda \). Obviously, these multiple Eves can be simplified to one Eve provided that a certain density and a certain area are given. Denote \( e \in \Omega \) as an arbitrary Eve. Using expectation by conditioning, the probability in (46) can be written as

\[ \Pr\{ \Gamma < \nu \} = \mathbb{E}_\Omega \left[ \prod_{e \in \Omega} \Pr\{ \Gamma_e < \nu | \Omega \} \right] \]  

(74a)

\[ = \mathbb{E}_\Omega \left[ \exp \left( \sum_{e \in \Omega} \log \left( \Pr\{ \Gamma_e < \nu | \Omega \} \right) \right) \right] . \]  

(74b)

In our scenario, each Bob assumes Eves are distributed according to the PPP \( \Omega \) in a circle around him with radius \( d_0 \). The relation between \( d_{qe} \) and \( d'_{qe} \) can be written as \( d_{qe} = \sqrt{d_{qq}^2 + d'_{qe}^2 - 2d_{qq}d'_{qe} \cos \varphi} \), where \( \varphi \) is the angle between \( d'_{qe} \) and \( d_{qq} \) that is uniformly distributed in the range \([0, 2\pi]\). Thus,

\[ \Gamma = \left( \frac{\beta}{\sqrt{d_{qq}^2 + \beta^2 - 2d_{qq} \beta \cos \varphi}} \right)^\eta . \]  

(75)
Let \( d'_{qe} = \beta \). The expectation in (74b) is equivalent to Laplace functional of a point process, so (74a) can be reduced to [33, Ch. 7]

\[
\Pr\{\Gamma_{\gamma} < \nu\} = \exp\left(-\lambda \int_0^{d_0} \int_0^{2\pi} \Pr\left\{ \left( \frac{\beta}{\sqrt{d_{qq}^2 + \beta^2 - 2d_{qq}\beta \cos \phi}} \right)^\eta > \nu \right\} \beta d\beta d\phi \right). \tag{76}
\]

Let \( \xi_q \triangleq \left( \frac{\beta}{\sqrt{d_{qq}^2 + \beta^2 - 2d_{qq}\beta \cos \phi}} \right)^\eta \), \( q \in Q \). The quantity \( \gamma \) in (76) is the SINR of a one-branch diversity combiner with \( N_q \) interferers whose CDF is [39]

\[
F_X(\gamma) = 1 - \frac{1}{1 + \gamma}. \tag{77}
\]

Using (77) in (76), we end up with

\[
\Pr\{\xi_q > \nu\} = (1 + \frac{|d'_{qq}H_{qq}|^2}{d_{qq}^2|d'_{qq}H_{qq}|^2})^{-N_q}. \tag{78}
\]

APPENDIX D

COMPARISON OF COMPLEXITY AND SIGNALING OVERHEAD BETWEEN MRC AND MMSE RECEIVERS

We first would like to mention that our work may not be applicable to devices with low computing capabilities like sensors or IoT devices. Note that these devices likely do not have multiple antennas anyway. However, our solution is in fact meant/designed for multiple antenna systems, e.g., smart phone, laptops, BSs whose computing powers are reasonably strong. We also avoided imposing additional computations on nodes in our proposed algorithms. For example, in the linear receiver stage, we chose MRC receivers instead of the MMSE receivers, as the MMSE method poses negligible performance improvement at the cost of additional complexity. In the following, a brief comparison between MMSE and MRC receivers in terms of number of operations is given\(^{35}\).

The first step to compute an MMSE receiver at each Bob is to calculate the covariance matrix of the interference at the receive chain of each Bob, which is basically a vector multiplication operation. Then, each Bob needs to measure the channel between himself and his corresponding

\(^{35}\)We skipped the detailed description of an MMSE method of reception for the sake of brevity. The fundamentals of MMSE receivers can be found at [22, chapter 6]. For more comparison between MRC and MMSE receivers, please see our response to comment 3 of reviewer 1, or see the simulation section of our manuscript.
Alice and multiply it to the inverse of the interference’s covariance matrix to establish the MMSE receiver. Compared to the MRC receiver, which only requires the channel between Bob and his corresponding Alice, the MMSE receiver requires three more operations (two matrix multiplications and one inverse) to be established, which can be significant depending on the number of Alice’s/Bob’s antennas.

Regarding the calculation of TxFJ and RxFJ powers, we first need to reintroduce the following definitions. The secrecy rate of Alice\(_q\) is denoted as \(C_{q}^{\text{sec}}\), and can be defined as

\[
C_{q}^{\text{sec}} \triangleq \max\{C_{q} - C_{eq}, 0\}
\tag{79}
\]

where \(C_{q}\) and \(C_{eq}\) are the information rate at Bob\(_q\) and the leaked rate at Eve from Alice\(_q\), respectively. \(C_{q}\) is defined as

\[
C_{q} \triangleq \log(1 + \frac{\phi_{q}P_{q}}{a_{q} + b_{q}p'_{q}})
\tag{80}
\]

where \(\phi_{q}\) is the power assignment (PA) for the information signal at Alice\(_q\), \(P_{q}\) is Alice\(_q\)’s transmit power, \(a_{q}\) is the normalized multi-user interference received at Bob\(_q\), \(b_{q}\) is the normalized self-interference channel at Bob\(_q\) and \(p'_{q}\) is the power of RxFJ. \(C_{eq}\) is defined as

\[
C_{eq} \triangleq \log(1 + \frac{\phi_{q}P_{q}}{c_{q} + d_{q}p'_{q}})
\tag{81}
\]

where \(c_{q}\) is the normalized interference received at Eve (except the interference received from the RxFJ of Bob\(_q\)), \(d_{q}\) is the normalized interference received from RxFJ of Bob\(_q\). To analyze the complexity of our power allocation algorithm, we first focus on the case where full knowledge of E-CSI is available\(^{36}\).

A. Computing the Optimal RxFJ Power

It was shown in the paper that the optimal value of RxFJ can be derived as follows:

\[
p'_{q}^{\ast} = \begin{cases} 
  P'_{q}, & \text{if } b_{q} < d_{q} \\
  0, & \text{if } b_{q} > d_{q}.
\end{cases}
\tag{82}
\]

where \(P'_{q}\) is the total power available at Bob\(_q\) for RxFJ. It can be seen from (82) that setting the optimal amount of RxFJ only involves a comparator to judge on the values of \(b_{q}\) and \(d_{q}\). Again,

\(^{36}\)We need to emphasize that we use the full-ECSI scenario to build foundation for our scheme to handle the case where knowledge of E-CSI is not available. The procedure for designing the scheme that is robust to E-CSI uncertainties is given in Section V.
we assume that in the full-ECSI scenario Alice, q knows the channel between herself and Eve; Moreover, MUI at Eve (i.e., \(c_q + d_q p_q^*\)) is also known at Alice, \(q \in Q\).

B. Computing the Optimal Power Allocation between Information and TxFJ Signals

The optimal PA (i.e., \(\phi_q, \forall q\)) can be found from the following optimization problem\(^{37}\):

\[
\begin{align*}
\text{maximize} & \quad C_{q}^{sec} \\
\text{s.t.} & \quad c_q = a_q + (b_q - d_q)p_q^* + \delta \\
& \quad c_q > 0 \\
& \quad 0 < \delta < (d_q - b_q)P_q^r + J(1 - t_q) \\
& \quad 0 \leq \phi_q \leq 1.
\end{align*}
\]

The optimal solution for PA (i.e., \(\phi_q\)) can be found by simplifying the following equality:

\[c_q = a_q + (b_q - d_q)p_q^* + \delta^*.\]

Notice that the term \(c_q\) includes \(\phi_q\), i.e.,

\[c_q = \frac{r_q^\dagger G_j q}{|r_q^\dagger G_q|^2} \sigma_q + O\]

where \(|r_q^\dagger G_j q|\) and \(|r_q^\dagger G_q|^2\) are the E-CSI components, \(\sigma_q = (1 - \phi_q)P_q\) is the power allocated to TxFJ and \(O\) covers other interference terms at Eve (see Eq. (9) of the paper). Hence, the simplification of the above equality w.r.t. \(\phi_q\) can be easily done. Note that (29) was derived only to proceed with the game-theoretic analysis of the problem. A detailed procedure to find the optimal value of \(\phi_q\) in a node is as follows.

At a given iteration of our algorithm, say the \(n\)th iteration, after setting the optimal value of RxFJ, in order to determine the optimal PA, Bob, q needs to first measure the interference at his receive chain, i.e., \(a_q^{(n-1)} + b_q^{(n-1)}p_q^*\) must be measured, where \(a_q^{(n-1)}\) and \(b_q^{(n-1)}\) indicate the values of \(a_q\) and \(b_q\) at the previous iteration. Assuming that full knowledge of E-CSI is available, Bob, q also knows the MUI at Eve in the previous iteration, i.e., \(c_q^{(n-1)} + d_q^{(n-1)}p_q^*\) is known\(^{38}\).

\(^{37}\)We skip details of deriving such optimization problem for the sake of brevity, as they already exist in the paper.

\(^{38}\)Notice that throughout the iterations of our algorithm, \(b_q^{(n-1)} = b_q^{(n)}\) and \(a_q^{(n-1)} = d_q^{(n)}\). However, the values of \(a_q\) and \(c_q\) can vary across iterations.
Hence, Bob_\textsubscript{q} does the following: 1) He subtracts the term \( \frac{|r_{q}^{j}G_{q}|c_{q}^{(n-1)}}{|r_{q}^{j}G_{q}|^2} \) from \( c_{q}^{(n-1)} \); 2) He adds the result of subtraction to \( d_{q}^{(n-1)}p_{q}^{*} \). Denote the result of this addition as \( g_{q} \); 3) He finds the optimal PA in the \( n \)th iteration, which can be described as:

\[
\phi_{q}^{*} = \max \left\{ \min \left\{ 1 - \frac{|r_{q}^{j}G_{q}|^2}{|r_{q}^{j}G_{q}|^2} (a_{q}^{(n-1)} + b_{q}^{(n-1)}p_{q}^{*} - g_{q}), 1 \right\}, 0 \right\}. \tag{86}
\]

It can be seen that setting the optimal PA involves simple addition, subtraction and division of scalar values. Moreover, there is no need to know all interference terms at Bob_\textsubscript{q} and Eve, only the aggregate of these terms (i.e., \( a_{q} \) and \( c_{q} \)) needs to be known. Knowing the noise floor at Bob_\textsubscript{q} can be helpful to measure the interference level. For example, in 802.11 systems, the noise level usually stays at \(-90 \text{ dBm} \) [32]. The computation of PA when E-CSI and MUI at Eve are not known still involves simple scalar operations, but is different in terms of the signaling it needs, i.e., the channels that Bob_\textsubscript{q} needs to know for his computations are different from the full-ECSI scenario. The signaling overhead associated to the robust scheme was already detailed in Section V.B of the current manuscript.

Thus, to the best of our knowledge, users with reasonably high computational capability can still perform the operations required by our algorithms with modest complexity.

**APPENDIX E**

**DETAILED ANALYSIS OF THE ROBUST SCHEME**

Note that we focus on no E-CSI knowledge in only Section V. However, for the purpose of laying a theoretical foundation, until Section V, we assume that E-CSI is available. In the scenario where knowledge of E-CSI is not available, we are in fact focused on optimizing the **ergodic secrecy rate**. In what follows, we give the details of our robust scheme. We first present a detailed formulation of our robust scheme to show that our robust scheme focuses on optimizing ergodic secrecy rate. Then, we provide the proofs of existence of NE as well as conditions that guarantee it uniqueness.
A. Detailed Formulation of the Robust Scheme

We first need to revisit the main optimization problem in (26). We have

\[
\max_{\phi_q, \delta} U_q(\phi_q, \delta, \xi) = C_{sq}^{sec}
\]

s.t. \[
c_q(\xi) = a_q + (b_q - d_q(\xi))p^{*}_q + \delta \\
c_q(\xi) > 0 \\
0 < \delta < (d_q(\xi) - b_q)P_q' + J(1 - t_q(\xi)) \\
0 \leq \phi_q \leq 1.
\]

where \(\xi\) is a parameter that indicates all E-CSI components. Note that the terms \(c_q, d_q\) and \(t_q\) are shown as functions of \(\xi\), as they depend on E-CSI components. When knowledge of E-CSI is not available, the parameter \(\xi\) can be treated as a random variable, i.e., \(\xi\) represents a random variable that maps the elements of a (continuous) set of random events \(\Omega\) to a real-valued vector which is referred to as E-CSI components. Now, we should optimize the expected value of \(U_q\) w.r.t. E-CSI components (i.e., \(\mathbb{E}_{\xi}[U_q(\phi_q, \delta, \xi)]\)), which is the same as optimizing the ergodic secrecy rate, (i.e., \(\mathbb{E}_{\xi}[C_{sq}^{sec}]\)). However, taking the expected value of the objective in (87) is not enough to convert problem (87) into a stochastic programming problem because the constraints of (87) also depend on E-CSI components. Without loss of generality, let \(\Omega\) be a set of infinitely many discrete events \(\omega_i, \ i = 1, 2, \ldots\), which are mapped to random variables \(\xi_i, \ i = 1, 2, \ldots\). Thus, the stochastic programming formulation of (87) can be written as [33]

\[
\max_{\phi_q, \delta} \mathbb{E}_{\xi}[U_q(\phi_q, \delta, \xi)] = \sum_{i=1} \Pr(\xi_i)U_q(\phi_q, \delta, \xi_i)
\]

s.t. \[
0 \leq \phi_q \leq 1. \\
c_q(\xi_i) = a_q + (b_q - d_q(\xi_i))p^{*}_q + \delta \\
c_q(\xi_i) > 0 \quad \forall i = 1, 2, \ldots \]

\[
0 < \delta < (d_q(\xi_i) - b_q)P_q' + J(1 - t_q(\xi_i)) \]

(88)
Removing the slack variable \( \delta \) gives us the following formulation:

\[
\begin{align*}
\max_{\phi_q} & \sum_{i=1}^{\Pr(\xi_i)} U_q(\phi_q, \xi_i) \\
\text{s.t.} & 0 \leq \phi_q \leq 1. \\
& c_q(\xi_i) \geq a_q + (b_q - d_q(\xi_i))p_q^* \\
& c_q(\xi_i) > 0 \quad \forall i = 1, 2, \ldots .
\end{align*}
\]  
(89)

Notice that in (88), the constraints need to hold for all \( \xi_i \) (i.e., all realization of E-CSI). In the jargon of stochastic programming, the first constraint in (89) is known as first-stage constraints, and the set of constraints that depend on \( \xi_i \) are referred to as second-stage constraints. In the case of finite set of random events (i.e., finite realizations of \( \xi_i \)) or some special types of objective functions, one can use two-stage stochastic programming approaches to efficiently solve (89) [33]. However, the set of random events is not finite in our case because the variations of the wireless environment are usually modeled as continuous distributions. Thus, the formulation in (89) becomes prohibitively difficult to solve with two-stage stochastic programming approaches. Therefore, we need to settle with a sub-optimal solution that is easier to achieve. To do this, we look at (89) again.

Recall that we already explained that the first second-stage constraint is mainly to do with allocating enough power to TxFJ to achieve positive secrecy. Ensuring that this constraint is satisfied across all \( \xi_i \) (i.e., all realizations of E-CSI) can be limiting. For example, for some (less probable) realizations of E-CSI, the channel between Alice\(_q\) and Eve can be a lot stronger than that between Alice\(_q\) and Bob\(_q\). Thus, ensuring positive secrecy for this realization can force Alice\(_q\) to allocate most of her power to TxFJ, which may be too conservative. We aim to avoid this issue by ensuring positive secrecy with a certain probability, i.e., positive secrecy is ensured across a subset of E-CSI realizations. In other words, the \( q \)th link needs to satisfy the following:

\[
\Pr\left((c_q(\xi_i) \geq a_q + (b_q - d_q(\xi_i))p_q^*)\right) \geq \varepsilon
\]  
(90)

where \( \varepsilon \) is a given probability for ensuring positive secrecy. An equivalent formulation of (90) already exists in the paper with different notations (see Section V). The probabilistic formulation in (90) later allowed us in the paper to find out a PA scheme for the case where E-CSI is unknown (see Section V). We also showed in simulations that a high value of \( \varepsilon \) although decreases the
leaked rate at Eve, it does not leave much power for the information signals and much freedom in interference management between legitimate links (see Section VI).

Overall, we saw that the main objective in our stochastic programming formulation was to maximize the ergodic secrecy rate. However, because of complexity of the stochastic programming formulation, we had to relax the stochastic programming formulation, which led us to the robust scheme we proposed in the paper.

The last issue is related to the on-off scheme proposed earlier for the RxFJ. As it was shown in Section III, the choice of RxFJ depends on whether \( b_q < d_q \) or \( b_q > d_q \). When \( b_q < d_q \) becomes a random variable in the case of unknown E-CSI, we choose to use RxFJ whenever \( \Pr\{b_q < d_q\} > 0.5 \). Computing \( \Pr\{b_q < d_q\} \) was detailed in Theorem 3.

### B. Existence and Uniqueness of NE in the Robust Scheme

We previously showed that in our robust scheme, the \( q \)th link, \( q \in Q \) aims satisfy the inequality in (40) (or equivalently in (90)), which leads to the following solution for the PA between information and FJ signals.

\[
\phi_q = \max \left\{ \min \left\{ 1 - \Pr\{b_q < d_q\} \frac{E[\psi_q + \tau_q P_q^e q]}{(1-\varepsilon)P_q} - \Pr\{b_q > d_q\} \frac{E[\psi_q]}{(1-\varepsilon)P_q}, 1 \right\}, 0 \right\}.
\]  

(91)

Notice that because our aim is to satisfy (90), then the stochastic program in (89) together with the set of constraints of (89) that hold (as a result of satisfying (90)) constitute a convex program with a compact and convex set of constraints. Hence, by using this stochastic formulation for each link, we can make sure that the NE exists in the robust scheme [30]. Now, we need to integrate the best responses of all links into one vector equation, so that we can establish a fixed-point problem and use the concepts of fixed-point theory to comment on the uniqueness of NE. Same as what we did in the proof of Theorem 1, we concatenate the solution in (91) for all \( q \) to establish the following fixed point problem in its \( t \)-th iteration

\[
\Phi^{(t+1)} = \mathbb{1} + \frac{1}{1-\varepsilon} \left( E[A + B] \Phi^{(t)} + E[f] \right)
\]  

(92)

where \( \Phi = [\phi_1, \ldots, \phi_Q]^T \) is the vector of PAs for all links, \( \mathbb{1} \) is a vector of appropriate size whose entries are all 1, and \( E[\bullet] \) is the expectation operator. It can be seen that (92) is similar to (36) with the only difference that in (92) we applied expectation w.r.t E-CSI to all terms. To analyze the uniqueness of NE, the fixed-point problem in (92) must be in closed form, i.e., the
expectation terms in (92) must be computable. Section V gives the close-form representation of these terms (see (45)–(52) and Theorem 3). Hence, the closed-form representation of links’ best responses can be derived, and all the NE uniqueness analysis that we did for the full-ECSI scenario in Theorem 1 is applicable in the robust scheme as well.

Using the robust formulation in Section V, the inequality in (53) (which gives us the conditions on the uniqueness of NE in the robust scheme) can be deduced the same way that we introduced (33) of the paper (to judge on the uniqueness of NE in the full-ECSI scenario). The difference is that instead of matrix $A + B$, we need to analyze the matrix $\frac{1}{1-\epsilon} E[A + B]$, where the expectation is element-wise and is w.r.t. ECSI components. Moreover, the same way that we claimed (and proved) (38) is a sufficient condition to satisfy (33) in the full-ECSI case, (54) can also be considered as a sufficient condition to satisfy (53).

REFERENCES


