Energy Harvesting for Secure OFDMA Systems

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Abstract—Energy harvesting and physical-layer security in wireless networks are of great significance. In this paper, we study the simultaneous wireless information and power transfer (SWIPT) in downlink orthogonal frequency-division multiple access (OFDMA) systems, where each user applies power splitting to coordinate the energy harvesting and information decoding processes while secrecy information requirement is guaranteed. The problem is formulated to maximize the aggregate harvested power at the users while satisfying secrecy rate requirements of all users by subcarrier allocation and the optimal power splitting ratio selection. Due to the NP-hardness of the problem, we propose an efficient iterative algorithm. The numerical results show that the proposed method outperforms conventional methods.

I. INTRODUCTION

Security is a significant issue in designing and developing wireless systems. Cryptography encryption as a traditional method dominates the upper layers by means of increasing the computational complexity. Security approaches have been introduced in every layer but physical layer in the standard fivelayered protocol stack. Thus physical-layer security has been an important complement to the other security approaches.

A great deal of studies have been devoted to informationtheoretic physical-layer security [1]–[5], In [1], [2], resource allocation for physical-layer security considerations was studied for multicarrier and multiple input multiple output (MIMO) systems. In [3], artificial noise was introduced in physicallayer security to obtain better performance in symmetric Gaussian interference channel. The authors in [4] studied the cooperative power control by using artificial noise in symmetric Gaussian interference channel. In [5], the authors studied the energy-efficient for physical-layer security in multicarrier systems.

On the other hand, simultaneous wireless information and power transfer (SWIPT) becomes an appealing solution to prolong the lifetime of wireless network nodes. It has drawn a great deal of research interests [6]–[10]. Energy-harvesting wireless networks are potentially able to gain energy from the wireless environment. The prior work [11]–[13] studied the performance of transmitting information and energy simultaneously in the receiver that can decode information and harvest energy for the same received signal, which may not be realizable however. Two practical schemes, so-called time switching and power splitting, were proposed in [7], [10], as practical designs. With time switching adopted at a receiver, the received signal is either processed for energy harvesting or for information decoding. When the power splitting applied at the receiver, the received signal can be split into two streams with one stream processed by the power harvester and the other processed by the information receiver. Flat-fading channel variations in SWIPT was studied in [8] and [9], where the receiver performs dynamic time switching or dynamic power splitting in the systems, respectively.

Orthogonal frequency division multiplexing access (OFDMA) gains its popularity and has become a leading technology in future broadband wireless networks, due to its flexibility in resource allocation and robustness against multipath. It enables efficient transmission of various data traffic by optimizing power, subcarrier, or bit allocation among different users. Plenty of works have taken into account either physical-layer security or SWIPT issue in OFDMA, such as [5], [10], [14], [15]. For instance, energy-efficient resource allocation for physical-layer security with multiple antennas and artificial noise was investigated in [5]. Resource allocation with SWIPT in OFDMA with different configurations was considered in [15].

The combination of SWIPT and private message exchanges between mobiles users and base station (BS) can be promising, since it not only meets mobiles users' secrecy requirements and but also enables users to harvest energy. In an OFDMA system, if a specific subcarrier is unsatisfactory for one user to transmit information, to transfer power instead can achieve a high efficient goal and prolong the lifetime of wireless networks. However, SWIPT based OFDMA system with security considerations has not been considered yet in the literature. This motivates us to investigate physical-layer security in SWIPT based OFDMA systems.

In [16], a case of power splitting applied at each subcarrier was considered. In practical circuits, power splitting is performed before OFDM demodulation. As a result, power splitting should be considered to perform on each user. However, the complexity of splitting on all subcarriers with the same ratio is much greater than splitting on each subcarrier with dynamic ratio. Furthermore, the Lagrangian function cannot be decomposed at each subcarrier. As a result, the optimal solution for power splitting at each user is difficult to obtain.

In this study, we consider the secrecy-rate required multiusers in downlink OFDMA networks, where the users apply

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power splitting to coordinate energy harvesting and information decoding processes. Our goal is to find efficient subcarrier allocation policies with power splitting ratio to maximize the aggregate harvested energy of all users satisfying the secrecy rate constraints among all users.

First, we formulate a problem as power splitting applied at each receiver for practical application (P-PA). We next formulate a problem as performance upper bound (P-UB) by assuming that power splitting is able to be applied at each subcarrier. Both problems are formulated as mixed integer programming problems and NP hard. For Problem (P-UB), the optimal solution is obtained by dual method. For Problem (P-PA), we propose an efficient iterative algorithm to find the solution that is very close to the performance upper bound.

II. SYSTEM MODEL AND PROBLEMS FORMULATION

In this paper, we consider the downlink of an OFDMA network with one BS, K mobile users, over N subcarriers. Each user communicates with BS and demands a secrecy rate that is no lower than a constant C_k , for all $1 \le k \le K$. Here we assume that equal power allocation is performed by the BS over all subcarriers for simplicity. This is reasonable since the gain brought by power adaption is very limited in OFDMA systems [17]–[21]. The receivers are considered to split the received signal into two signal streams, with one stream to the energy receiver and the other one to information receiver. Furthermore, we assume for each user, any other user in the same network is a potential eavesdropper. The BS can obtain full statistical knowledge and instantaneous knowledge of CSI and each subcarrier is occupied only by one user at each time. This is reasonable since all users are assumed to be legitimate users and have their own data transmission with the BS.

Let $h_{k,n}$ denote the channel gain of user k on subcarrier n, and $\beta_{k,n}$ denote the channel gain of the potential eavesdropper for user k on subcarrier n, i.e., $\beta_{k,n} = \max_{k',k'\neq k} h_{k',n}$. We use a binary assignment variable $x_{k,n}$ to represent the subcarrier allocation, with $x_{k,n} = 1$ indicating that subcarrier n is assigned to user k and $x_{k,n} = 0$ otherwise. Let p_n represent the equal power allocated on subcarrier n.

All subcarriers have to be power split from the RF signals. Energy harvesting has to perform in analog domain instead of the digital domain where information of each subcarrier is decoded. As a result, due to the hardware limitation, for each mobile station, the received signal has to be harvested with a same power splitting ratio on all subcarriers.

We first consider the practical scenario, that is the received signal at user k is processed by a power splitter, where a ratio ρ_k of power is split to energy receiver and remaining ratio $1 - \rho_k$ of power is split into the information decoder, with $0 \le \rho_k \le 1$, $\forall k$. Thus, the achievable secrecy rate at subcarrier n when assigned to user k is

$$r_{k,n}^{s} = \left[\log\left(1 + \frac{(1-\rho_{k})p_{n}h_{k,n}}{\sigma^{2}}\right) - \log\left(1 + \frac{p_{n}\beta_{k,n}}{\sigma^{2}}\right)\right]^{+}$$
$$= \left[\log\left(\frac{(1-\rho_{k})p_{n}h_{k,n} + \sigma^{2}}{p_{n}\beta_{k,n} + \sigma^{2}}\right)\right]^{+}, \tag{1}$$

where $[\cdot]^+ = \max\{\cdot, 0\}.$

Then the secrecy rate of user k is denoted as

$$r_k^s = \sum_{n=1}^N x_{k,n} r_{k,n}^s.$$
 (2)

Assuming that the conversion efficiency of the energy harvesting process at each receiver is denotes by $0 < \zeta < 1$, the harvested power of user k is thus given by

$$E_k = \zeta \rho_k \sum_{n=1}^N p_n h_{k,n}.$$
(3)

The goal of the considered problem is to find the optimal subcarrier allocation and power splitting ratio in order to maximize the total harvested power (for the purpose of uplink transmission for example) while satisfying the individual secrecy rate requirement for each user. This practical application optimization problem can thus be expressed as

$$(P - PA): \max_{\{\mathbf{X}, \boldsymbol{\rho}\}} \qquad \zeta \sum_{k=1}^{K} \rho_k \sum_{n=1}^{N} p_n h_{k,n} \tag{4}$$

s.t.
$$\sum_{k=1}^{n} x_{k,n} \le 1, \forall n \tag{5}$$

$$x_{k,n} \in \{0,1\}, \forall k,n$$
 (6)

$$0 \le \rho_k \le 1, \forall k \tag{7}$$

$$\sum_{n=1} x_{k,n} r_{k,n}^s \ge C_k, \forall k \qquad (8)$$

where $X = \{x_{k,n}\}$, and $\rho = \{\rho_k\}$. The constraints in (5) and (6) enforce that each subcarrier can only be used by one user to avoid multi-user interference.

An upper bound for this problem can be obtained by assuming the power splitting is applied in receiver to split power for each subcarrier¹. In this case, ρ_k is extended to $\rho_{k,n}$. Thus, we consider the following optimization problem as

$$(P - UB): \max_{\{\boldsymbol{X}, \boldsymbol{\rho}\}} \qquad \zeta \sum_{k=1}^{K} \sum_{n=1}^{N} \rho_{k,n} p_n h_{k,n} \qquad (9)$$

$$0 \le \rho_{k,n} \le 1, \forall k, n \tag{10}$$

where

$$r_{k,n}^{s} = \left[\log \left(\frac{(1 - \rho_{k,n}) \, p_n h_{k,n} + \sigma^2}{p_n \beta_{k,n} + \sigma^2} \right) \right]^+.$$
(11)

¹This is because of the improved flexibility of resource allocation. In this case, users can harvest all the power at the subcarriers that are not assigned to them and thus improve the harvested power, i.e., $\rho_{k,n} = 1$ for all $x_{k,n} = 0$.

III. OPTIMAL POWER SPLITTING RATIO SELECTION AND SUBCARRIER ALLOCATION FOR PRACTICAL APPLICATION

The formulated problem (P-PA) is non-convex, finding the optimal solution is usually prohibitively due to the complexity. However, according to [22], the duality gap becomes zero in multicarrier systems as the number of subcarriers goes infinity for satisfying time-sharing condition.

We define \mathcal{T} as all sets of possible X that satisfy (5) and (6), \mathcal{R} as all sets of possible ρ for given X that satisfy $0 \le \rho_k \le 1$ for $x_{k,n} = 1$ and $\rho_k = 1$ for $x_{k,n} = 0$.

The Lagrangian function for this problem is given as follow:

$$L(\boldsymbol{\rho}, \boldsymbol{X}, \boldsymbol{\mu}) = \zeta \sum_{k=1}^{K} \rho_k \sum_{n=1}^{N} p_n h_{k,n} + \sum_{k=1}^{K} \mu_k \left(\sum_{n=1}^{N} x_{k,n} r_{k,n}^s - C_k \right)$$
$$= \sum_{k=1}^{K} \sum_{n=1}^{N} \left(\zeta \rho_k p_n h_{k,n} + x_{k,n} \mu_k r_{k,n}^s \right) - \sum_{k=1}^{K} \mu_k C_k, \quad (12)$$

where $\boldsymbol{\mu} = [\mu_1, \mu_2, ..., \mu_k]^T$ are the Lagrange multipliers. The dual function is then defined as

$$g(\boldsymbol{\mu}) = \max_{\boldsymbol{X} \in \mathcal{T}, \boldsymbol{\rho} \in \mathcal{R}(\boldsymbol{X})} L(\boldsymbol{\rho}, \boldsymbol{X}, \boldsymbol{\mu}).$$
(13)

Then the dual problem is thus given by $\min_{\mu \geq 0} g(\mu)$.

Through simple observation, when considering the maximization problem in (13). The Lagrangian function cannot be decomposed into N subproblems. Because optimal ρ_k^* has to be computed considering all subcarriers that are assigned to the kth user, instead of only considering one specific subcarrier.

Thus, we will obtain a suboptimal solution by iteratively optimizing X with a fixed set of ρ , and then optimizing ρ with a fixed set of X, which is known as block-coordinate descent method [23].

A. Optimality Condition of Subcarrier Assignment

First, we notice that the problem in (13) can be decomposed into N subproblems at each subcarrier in order to solve the X with a given set of ρ as

$$\max_{\boldsymbol{X}_n \in \mathcal{T}} L_n(\boldsymbol{X}_n) = \zeta p_n \sum_{k=1}^K \rho_k h_{k,n} + \mu_k r_{k,n}^s, \qquad (14)$$

which can be solved independently.

Note that for the Lagrangian (12), we have

$$L = \sum_{n=1}^{N} L_n - \sum_{k=1}^{K} \mu_k C_k.$$
 (15)

As a result, the optimal $x_{k,n}$ can be obtained as

$$x_{k,n}^* = \begin{cases} 1, \text{ if } k = k^* = \arg\max_k \mathcal{X}_{k,n} \\ 0, \text{ otherwise,} \end{cases}$$
(16)

where $\mathcal{X}_{k,n} = \zeta p_n \sum_{k=1}^{K} \rho_k h_{k,n} + \mu_k r_{k,n}^s$.

B. Optimal Power Splitting Ratio

Next, we consider the problem in (13) with a given set of feasible X. The problem thus can be decomposed into K subproblems and each for one user, which can be solved independently. This subproblems at user k is

$$\max_{\rho_k \in \mathcal{R}(\mathbf{X})} \qquad L_k(\rho_k) = \sum_{n=1}^N \left(\zeta \rho_k p_n h_{k,n} + x_{k,n} \mu_k r_{k,n}^s \right),$$
(17)

where

$$L = \sum_{k=1}^{K} L_k - \sum_{k=1}^{K} \mu_k C_k.$$
 (18)

According to Karush-Kuhn-Tucker (KKT) conditions, the optimal ρ_k^* has to satisfy the following equation:

$$\frac{\partial L_k}{\partial \rho_k} = \sum_{n=1}^{N} \left[\zeta p_n h_{k,n} - \frac{\mu_k x_{k,n} h_{k,n} p_n}{\ln 2 \left(h_{k,n} p_{k,n} \left(1 - \rho_k \right) + \sigma^2 \right)} \right] = 0.$$
(19)

According to (19), there is no closed-form for the optimal ρ_k^* . However $\frac{\partial L_k}{\partial \rho_k}$ monotonically decreases as ρ_k increase. Thus a bisection search method for optimal ρ_k^* is feasible.

With the fixed ρ , the optimal X can be obtained by (16). The optimal value of Problem (P-PA) can be increased by optimizing X. Thus the above process can be iterated until the optimal value of problem ceases to improve. In this case, the updated subcarrier allocation variables X remain the same as they are before the update. Since We notice that the local optimal solution depends on the initial sets of ρ , M sets of ρ are initialized to obtain a robust solution. We obtain the optimal X for each initialization step of ρ and update the ρ by a bisection search so as to obtain the maximal Lagrangian, and the iterative algorithm is applied to obtain a local optimal solution. Then, the final solution is the one achieves the greatest harvested power.

In general, as the M increases, greater robustness and optimality of the algorithm can be guaranteed. As $M \rightarrow \infty$, this algorithm is considered to perform as well as the optimal algorithm. However, large number of initialization steps increases the computation complexity, which will not be favorable for practical applications.

C. Subgradient updating

Finally, in [24], the dual function in (13) is always convex. By simultaneously updating μ , we can solve this problem by subgradient method. The dual variables μ are updated in parallel as follow

$$\mu_{k}^{(t+1)} = \left[\mu_{k}^{(t)} + \alpha_{k} \left(C_{k} - \sum_{n=1}^{N} x_{k,n} r_{k,n}^{s}\right)\right]^{+}, \forall k. \quad (20)$$

To summarize, the above iterative algorithm to solve Problem (P-PA) is given in Algorithm 1. For this algorithm, the computational complexity mainly lies in the steps 2)-23).

Algorithm 1 Proposed Iterative Algorithm

- 1: initialize Randomly generate M feasible sets of X as different initialization steps.
- 2: for each set of initialization do

power from M solution.

- 3: initialize μ .
- 4: repeat
- 5: repeat
- 6: Compute $r_{k,n}^s$ for all k, and n according to (1).
- 7: Solve assignment variables X according to (16).
- 8: **for** each optimal ρ_k to achieve the maximal L_k **do**
- **initialize** $\rho_k^{UB} = 1$ and $\rho_k^{LB} = 0$, then $\rho_k = (1/2) \left(\rho_k^{UB} + \rho_k^{LB} \right)$. 9: repeat 10: Compute $\frac{\partial L_k}{\partial \rho_k}$ according to (19). if $\frac{\partial L_k}{\partial \rho_k} > 0$ then Set $\rho_k^{LB} = \rho_k$. 11: 12: 13: 14: else Set $\rho_k^{UB} = \rho_k$. 15: end if $\left|\frac{\partial L_k}{\partial \rho_k}\right| < \varepsilon.$ 16: 17: end for 18: until X converge. 19: Update μ by (20) according to ellipsoid method. 20: until μ converge. 21: 22: end for 23: Find the one which achieves the maximum harvested

As the ρ_k are obtained individually by bisection search, the complexity of steps 11)-20) is $\mathcal{O}(K)$. Hence, we can also obtain the complexity of steps 5)-21) is $\mathcal{O}(K + KN)$. Next, the complexity of subgradient updates is polynomial in K [24]. We have the computational complexity of steps 2)-23) is $\mathcal{O}(K^{q+1} + K^{q+1}N)$, where q is a constant. Finally, considering further the M initialization steps, the time complexity of this algorithm is $\mathcal{O}(K^{q+1}M + K^{q+1}NM)$.

IV. PERFORMANCE UPPER BOUND

Then we can derive the Lagrangian function of Problem (P-UB) as follows

$$L(\boldsymbol{X}, \boldsymbol{\rho}, \boldsymbol{\lambda}) = \sum_{k=1}^{K} \sum_{n=1}^{N} \zeta \rho_{k,n} p_n h_{k,n}$$
$$+ \sum_{k=1}^{K} \lambda_k \left(\sum_{n=1}^{N} x_{k,n} r_{k,n}^s - C_k \right)$$
$$= \sum_{k=1}^{K} \sum_{n=1}^{N} \left(\zeta \rho_{k,n} p_n h_{k,n} + x_{k,n} \lambda_k r_{k,n}^s \right) - \sum_{k=1}^{K} \lambda_k C_k, \quad (21)$$

where $\boldsymbol{\lambda} = [\lambda_1, \lambda_2, ..., \lambda_K]^T$ is the vector of dual variables. The Lagrangian dual function can be obtained as

$$g(\boldsymbol{\lambda}) = \max_{\boldsymbol{X} \in \mathcal{T}, \boldsymbol{\rho} \in \mathcal{R}(\boldsymbol{X})} L(\boldsymbol{X}, \boldsymbol{\rho}), \quad (22)$$

and we can obtain the dual problem as $\min_{\lambda \geq 0} g(\lambda)$.

The dual function $g(\lambda)$ can be decomposed into N subproblems which can be solved independently. The subproblems can be obtained as

$$\max_{\boldsymbol{X}\in\mathcal{T},\boldsymbol{\rho}\in\mathcal{R}(\boldsymbol{X})} L_n(\boldsymbol{X}_n,\boldsymbol{\rho}_n) = \zeta p_n \sum_{k=1}^K \rho_{k,n} h_{k,n} + \lambda_k r_{k,n}^s,$$
(23)

where

$$L = \sum_{n=1}^{N} L_n - \sum_{k=1}^{K} \lambda_k C_k.$$
 (24)

We first seek for the optimal power splitting ratio of each subcarrier. According the Karush-Kuhn-Tucker (KKT) conditions [24], we take the partial derivation of $L_n(\mathbf{X}_n, \boldsymbol{\rho}_n)$ with respect to $\rho_{k,n}$, and equate it to zero.

$$\frac{\partial L_n}{\partial \rho_{k,n}} = \sum_{k=1}^K \zeta p_n h_{k,n} - \frac{\lambda_k h_{k,n} p_n}{\ln 2 \left(h_{k,n} p_{k,n} \left(1 - \rho_{k,n} \right) + \sigma^2 \right)} = 0.$$
(25)

Note that $\frac{\partial r_{k,n}^s}{\partial \rho_{k,n}} = 0$ when $\rho_{k,n} \leq 1 - \beta_{k,n}/h_{k,n}$. Thus the above equation is only true when $\rho_{k,n} > 1 - \beta_{k,n}/h_{k,n}$.

When $\rho_{k,n} \leq 1 - \beta_{k,n}/h_{k,n}$, we have

$$\frac{\partial L_n}{\partial \rho_{k,n}} = \sum_{k=1}^K \zeta p_n h_{k,n},\tag{26}$$

which is always greater than zero.

Thus, the optimal solution $\rho_{k,n}^*$ for both cases can be readily given by

$$\rho_{k,n}^* = \begin{cases} \rho_{k,n}, \text{ if } \rho_{k,n} \leq 1 - \beta_{k,n}/h_{k,n} \\ 1, \text{ otherwise,} \end{cases}$$
(27)

where

$$\dot{\rho_{k,n}} = \left[1 - \frac{\lambda_k h_{k,n}}{\zeta \ln 2p_n \sum_{k=1}^K h_{k,n}} + \frac{\sigma^2}{\ln 2h_{k,n}p_n}\right]_0^1, \quad (28)$$

and $[\cdot]_a^b = \max\{\min\{\cdot, b\}, a\}.$

Next, substituting (27) into $L_n(\mathbf{X}_n, \boldsymbol{\rho}_n)$, the optimal subcarrier assignment policy is given by (the details are easy and omitted here).

$$x_{k,n}^* = \begin{cases} 1, \text{ if } k = k^* = \arg\max_k \mathcal{H}_{k,n} \\ 0, \text{ otherwise,} \end{cases}$$
(29)

where $\mathcal{H}_{k,n} = \sum_{k=1}^{K} \zeta \rho_{k,n}^* p_n h_{k,n} + \lambda_k r_{k,n}^s$. Then, dual variable λ can be updated as follow

$$\lambda_k^{(t+1)} = \left[\lambda_k^{(t)} + \alpha_k \left(C_k - \sum_{n=1}^N x_{k,n} r_{k,n}^s\right)\right]^\top, \forall k. \quad (30)$$

The complexity of the dual based algorithm is analyzed as follows. For each subcarrier, $\mathcal{O}(K)$ computations are needed. Since the calculation is independent at each subcarrier, the

Algorithm 2 Dual-based method Algorithm for Upper Bound

- 1: initialize λ .
- 2: repeat
- 3: Compute $\rho_{k,n}$ according to (27) and (28), and then $r_{k,n}^s$ according to (11) for all k and n.
- 4: Solves assignment variables $x_{k,n}$ according to (29).
- 5: Update λ via (30) according to ellipsoid method.
- 6: **until** λ converge.

complexity if $\mathcal{O}(KN)$ for each iteration. Last, considering the the complexity of subgradient updates, the overall complexity of subgradient method is $\mathcal{O}(K^{q+1}N)$, where q is a constant. Finally, we present the whole algorithm in Algorithm 2.

V. NUMERICAL RESULTS

In this section, we provide some numerical results to evaluate the performance of the proposed iterative algorithm. In the simulation setup, we consider an OFDMA network with N = 128 and K = 8 mobile users. The users are Rayleigh distributed between a reference distance d_0 and 10m, where $d_0 = 1$ m and results in -30 dB path loss. In addition, the power is allocated to each subcarrier uniformly, which can be also described as $p_n = P_t/N$, where P_t is the total transmit power of the BS. Let E_{sum} denote the overall power harvested by all users. For all energy receivers, it is assumed that $\zeta = 0.4$. The minimum secrecy rate \overline{C} is assumed for the first 4 of the users, i.e., $C_1 = C_2 = C_3 = C_4 = \overline{C}$. And for the rest 4 users, they have no requirement of secrecy rate, i.e., $C_5 = C_6 = C_7 = C_8 = 0$. For the information receiver, $\sigma^2 = -30$ dBm. In addition, for the proposed iterative algorithm, we set M = 200.

To evaluate the schemes, we introduce another two schemes in this simulation as benchmarks. In the first scheme, denoted as fixed power splitting (FPS), power splitting ratio $\rho_k = 0.5, \forall k$, is fixed for all users beforehand while Xare computed according to (29). In the second scheme, the subcarrier assignment is fixed (FSA), while ρ_k for each user is optimized by a bisection search as Algorithm 1.

We first demonstrate the pairs of achievable overall harvested power E_{sum} and feasible secrecy rate requirement \bar{C} . Fig. 1 shows the rate pair at fixed total transmit power $P_t = 15$ dBm. First, it is observed that for all schemes, E_{sum} decreases with the increase of secrecy requirement \bar{C} . In addition, for the proposed algorithm and upper bound, E_{sum} falls sharply to zero at around $\bar{C} = 10.5$ bit/OFDM symbol. It is observed that according to the performance of upper bound and the proposed algorithm, power splitting on each user only incurs a little in harvested power when achieving the same \bar{C} . On the other hand, the two algorithms earn great advantage over these two benchmarks FPS and FSA. In addition, the maximum feasible points of the FPS and FSA appear at around $\bar{C} = 0.5$ and 3.8 bit/OFDM symbol respectively, which are much lower than that in the proposed algorithm.

Fig. 2 show the overall power consumption for information receivers of all users versus \overline{C} . We notice that the proposed



Fig. 1. Achievable E_{sum} versus \overline{C} pair at total transmit power of 15 dBm.

algorithm only consumes a little bit more power than the upper bound when achieving the same \bar{C} . In addition, the power consumption for FSA and FPS increases sharply at around $\bar{C} = 0.5$ and 3.8 bit/OFDM symbol respectively.

Fig. 3 depicts the overall power consumption for information receivers versus the maximum transmit power P_t under $\bar{C} = 0.4$ bit/OFDM symbol. Both algorithms consume much less power than FPS and FSA schemes. FPS and FSA consume 300% power more than the proposed algorithm. Moreover, the proposed algorithm performs closely to the upper bound.

Finally, we demonstrate the relation between E_{sum} and total transmit power P_t under $\overline{C} = 0.4$ bit/OFDM symbol in Fig. 4. We can conclude that the proposed algorithm is also very close to the performance upper bound. At low SNR region $(P_t < 12 \text{dBm})$, all schemes perform close to each other. As P_t becomes larger, the proposed algorithm only incurs a marginal performances loss, and both significantly outperform the two benchmarks.

VI. CONCLUSIONS

This study investigated the joint subcarrier allocation policy and power splitting ratio selection for downlink security OFDMA broadband networks. We formulated the problem to maximize the harvested power while satisfying the secrecy rate requirements of all users. Simulation results show that the proposed algorithm performs closely to upper bound and significantly outperforms the conventional methods.

REFERENCES

- E. A. Jorswieck and A. Wolf, "Resource allocation for the wire-tap multicarrier broadcast channel," in *Proc. Int. Conf. Telecommun*, Junuary 2008, pp. 1–6.
- [2] Z. Li, R. Yates, and W. Trappe, "Secrecy capacity of independent parallel channels," in *Proc. 44th. Allerton Conf. Commun., Control Comput.,*, September 2006, pp. 841–848.
- [3] S. Goel and R. Negi, "Guaranteeing secrecy using artificial noise," *IEEE Trans. Wireless Commun.*, vol. 7, no. 6, pp. 2180–2189, Junuary 2008.



Fig. 2. Total power consumption for all information receivers versus \bar{C} pair at total transmit power of 15 dBm.



Fig. 3. Total power consumption for all information receivers versus P_t at $\bar{C} = 0.4$ bit/OFDM symbol.

- [4] J. Zhu, J. Mo, and M. Tao, "Cooperative secret communication with artificial noise in symmetric interference channel," *IEEE Commun. Lett.*, vol. 14, pp. 885–887, October 2010.
- [5] D. W. K. Ng, E. S. Lo, and R. Schober, "Energy-efficient resource allocation for secure OFDMA systems," *IEEE Trans. Veh. Technol.*, vol. 61, no. 6, pp. 2575–2585, July 2012.
- [6] D. Zhang, Y. Wang, and J. Lu, "QoS aware relay selection and subcarrier allocation in cooperative OFDMA systems," *IEEE Commun. Letters*, vol. 12, no. 4, pp. 294–296, 2012.
- [7] R. Zhang and C. K. Ho, "MIMO broadcasting for simultaneous wireless information and power transfer," *IEEE Trans. Wireless Commun.*, vol. 12, no. 5, pp. 1989–2001, May 2013.
- [8] L. Liu, R. Zhang, and K. Chua, "Wireless information transfer with opportunistic energy harvesting," *IEEE Trans. Wireless Commun.*, vol. 12, no. 1, pp. 288–300, January 2013.
- [9] L. Liu, R. Zhang, and K. C. Chua, "Wireless information and power transfer: a dynamic power splitting approach," *IEEE Trans. Commun.*, vol. 61, pp. 3990 – 4001, September 2013.



Fig. 4. Achievable E_{sum} versus P_t at $\overline{C} = 0.4$ bit/OFDM symbol.

- [10] X. Zhou, R. Zhang, and C. K. Ho, "Wireless information and power transfer in multiuser OFDM systems," *IEEE Trans. Wireless Commun.*, vol. 13, no. 4, pp. 2282–2294, 2014.
- [11] L. R. Varshney, "Transporting information and energy simultaneously," in *Proc. IEEE ISIT*, 2008.
- [12] P. Grover and A. Sahai, "Shannon meets Tesla: wireless information and power transfer," in *Proc. IEEE ISIT*, 2010.
- [13] A. Fouladgar and O. Simeone, "On the transfer of information and energy in multi-user systems," *IEEE Comm. Lett.*, vol. 16, pp. 1733– 1736, 2012.
- [14] X. Wang, M. Tao, J. Mo, and Y. Xu, "Power and subcarrier allocation for physical-layer security in OFDMA-based broadband wireless networks," *IEEE Trans. Inf. Forersics Security*, vol. 6, no. 3, pp. 693–702, September 2011.
- [15] K. Huang and E. Larsson, "Simultaneous information and power transfer for broadband wireless systems," *IEEE Trans. Signal Proc.*, vol. 61, no. 23, pp. 5972– 5986, December 2013.
- [16] D. W. K. Ng, E. S. Lo, and R. Schober, "Wireless information and power transfer: Energy effciency optimization in OFDMA systems," *IEEE Trans. Wireless Commun.*, vol. 12, pp. 6352–6370, December 2013.
- [17] Y. Liu, M. Tao, B. Li, and H. Shen, "Optimization framework and graph-based approach for relay-assisted bidirectional OFDMA cellular networks," *IEEE Trans. Wireless Commun.*, vol. 9, no. 11, pp. 3490– 3500, November 2010.
- [18] Y. Liu and M. Tao, "Optimal channel and relay assignment in OFDMbased multi-relay multi-pair two-way communication networks," *IEEE Trans. Commun.*, vol. 60, no. 2, pp. 317–321, Feburary 2012.
- [19] Y. Liu, J. Mo, and M. Tao, "QoS-aware transmission policies for OFDM bidirectional decode-and-forward relaying," *IEEE Trans. Wireless Commun.*, vol. 12, no. 5, pp. 2206–2216, 2013.
- [20] H. Zhang, Y. Liu, and M. Tao, "Resource allocation with subcarrier pairing in OFDMA two-way relay networks," *IEEE Wireless Commun. Lett.*, vol. 1, no. 2, pp. 61–64, Apr. 2012.
- [21] M. Tao and Y. Liu, "A network flow approach to throughput maximization in cooperative OFDMA networks," *IEEE Trans. Wireless Commun.*, vol. 12, no. 3, pp. 1138–1148, 2013.
- [22] W. Yu and R. Lui, "Dual methods for nonconvex spectrum optimization of multicarrier systems," *IEEE Trans. Commun.*, vol. 54, pp. 1310–1322, July 2006.
- [23] P. Richtarik and M. Takac, "Iteration complexity of randomized blockcoordinate descent methods for minimizing a composite function," *Mathematical Programming, Series A*, December 2012.
- [24] S. Boyd and L. Vandenberghe, *Convex Optimization*. Cambridge University Press, 2004.