

Resource Allocation for Device-to-Device Communications Underlying Heterogeneous Cellular Networks Using Coalitional Games

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Abstract—Heterogeneous cellular networks (HCNs) with millimeter wave (mm-wave) communications included are emerging as a promising candidate for the fifth generation mobile network. With highly directional antenna arrays, mm-wave links are able to provide several Gbps transmission rate. However, mm-wave links are easily blocked without line of sight. On the other hand, device to device (D2D) communications have been proposed to support many content-based applications and need to share resources with users in HCNs to improve spectral reuse and enhance system capacity. Consequently, an efficient resource allocation scheme for D2D pairs among both mm-wave and the cellular carrier band is needed. In this paper, we first formulate the problem of the resource allocation among mm-wave and the cellular band for multiple D2D pairs from the view point of game theory. Then, with the characteristics of cellular and mm-wave communications considered, we propose a coalition formation game to maximize the system sum rate in statistical average sense. We also theoretically prove that our proposed game converges to a Nash-stable equilibrium and further reaches the near-optimal solution with fast convergence rate. Through extensive simulations under various system parameters, we demonstrate the superior performance of our scheme in terms of the system sum rate compared with several other practical schemes.

Index Terms—Device-to-device communication, game theory, HCNs, millimeter wave communication, resource allocation.

I. INTRODUCTION

WITH the increasing proliferation of mobile devices with high capabilities and intelligence, the global mobile traffic is expected to experience a remarkable and continuous growth in the next few years. As predicted by Cisco, the traffic generated from wireless and mobile devices is expected to constitute a major percentage of the total internet protocol (IP) traffic by 2020. It is also estimated that the number of devices accessed to IP networks will be three times of the global population in 2020 and the mobile traffic will grow at an annual rate of 53% until 2020 [1]. At the same time, the millimeter wave (mm-wave) has huge bandwidth, and therefore, much higher network capacity can be achieved [2]. There are already several standards defined for indoor wireless personal area networks (WPANs) or wireless local area networks (WLANs) in the mm-wave band, such as ECMA-387 [3], IEEE 802.15.3c [4], and IEEE 802.11ad. Thus, in order to keep up with the explosive growth of mobile devices and data traffic, one key enabling solution is to exploit HCNs in both the cellular band and the mm-wave band.

HCNs operating in both conventional cellular band and in the mm-wave band, can improve the system performance effectively. Two kinds of networks offer different advantages. For example, cellular network provides higher link reliability, while mm-wave communication has obvious advantages in the transmission rate. However, a most common concern is that mm-wave communications suffer a much larger distance-dependent propagation loss due to the high carrier frequency [5], [6]. For example, the free space path loss at the 60 GHz band is 28 dB more than that at 2.4 GHz [7]. To combat severe channel attenuation, we utilize the highly directional antennas and the beamforming technology at both the transmitter and the receiver [8]. Moreover, mm-wave communication typically requires line of sight (LOS) communication.

Device to device (D2D) communications underlying the HCN, as a method of great potential to offload traffic from the base station (BS), can improve network performance and provide a better user experience [9], [10]. Under the coverage of

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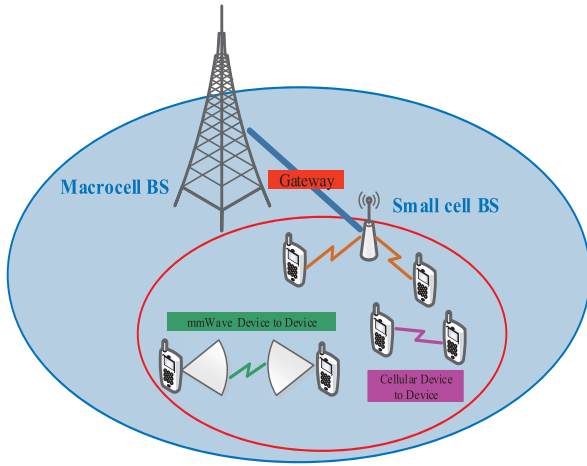


Fig. 1. The D2D-enabled HCN underlaying the macrocell.

BS, user equipments (UEs) in physical proximity communicate with each other directly using the resources in the mm-wave band or sharing resources with cellular users. The integration of D2D communications into HCNs has the advantage of allowing for high data rate, low delay and power consumption transmission for popular proximity-based applications. Consequently, these high quality D2D links generate the hop gain by transmitting data signals directly between two closely located terminals without involving a centralized controller. On the other hand, reuse gain is achieved by simultaneously using the same radio resource for cellular users and D2D pairs. Additionally, the D2D-enabled HCN also facilitates new types of peer-to-peer services.

In Fig. 1, we show a typical scenario of the D2D-enabled HCN underlaying the macrocell. Cellular users are associated with BS of the small cell, which is connected to the BS of the macrocell via the gateway. In the D2D-enabled HCN, interference produced by D2D communications hampers the performance of cellular communications. Intra-cell interference, which is referred to the interference between users as the result of spectrum sharing, is considered to be an important and complex problem in HCNs, especially the interference between D2D pairs [11]. Therefore, it is necessary to investigate and properly deal with the interference problems such that the benefits of proximity transmissions can be fully exploited. To date, extensive works have been undertaken on the power control [12]–[14], resource allocation [15]–[19] and association techniques among the cellular users and D2D pairs to mitigate the interference and obtain the maximum system achievable transmission rate. Besides, based on the differences between cellular D2D networks and mm-wave D2D networks, how to utilize the advantages of both networks to optimize the sub-channel allocation under HCNs indeed brings great challenges.

In this paper, we consider D2D communications in the HCN combining mm-wave and cellular networks for uplink resource allocation, and then formulate the problem of maximizing the system sum rate via resource allocation into a nonlinear integer programming problem. With the complicated interferences

considered among cellular users and D2D pairs, we address the problem of resource allocation for multiple cellular users and D2D pairs from a game theory point of view using coalition formation game [20]. The coalition game, which is widely used in wireless communications, for example, the resource allocation problems, allows several players cooperatively to form a coalition in order to optimize resource allocation, manage the interference, and further enhance the system performance. Then, we develop a coalition formation algorithm to achieve the Nash-stable equilibrium for the proposed coalition game. The main contributions of this paper can be summarized as follows.

- We introduce the coalition formation game to model the D2D communications underlaying HCN consisting of multiple cellular users and D2D pairs. Based on the established model, we investigate the resource allocation problems for the realistic HCNs.
- We formulate the problem of D2D resource allocation underlaying HCN aiming to enable massive connectivity and maximize the system sum rate. Then, we utilize the advantages of cellular D2D network and mm-wave D2D network, and develop a coalition formation algorithm to implement efficient resource allocation with low computation complexity. We show that the proposed algorithm converges to a Nash-stable coalition structure and achieves a near-optimal solution with fast convergence rate.
- Through extensive simulations under various system parameters, we evaluate the system performance of our proposed coalition game based approach compared with other practical schemes.

The rest of the paper is organized as follows. In Section II, we present an overview of the related work. Section III introduces the system model and formulates the resource allocation problem. The coalition game with transferable utility and corresponding algorithm is proposed in Section IV. We analyze the properties of the proposed algorithm in Section V. Section VI gives the performance evaluation of our proposed scheme compared with other schemes under various system parameters. Finally, the conclusions of this paper are drawn in Section VII.

II. RELATED WORK

There have been several related works studying resource allocation and interference management for D2D communications. For example, Ramezani-Kebrya *et al.* [12] proposed an efficient power control algorithm and jointly optimized the power of a cellular user and a D2D pair aiming at maximizing their sum rate, while providing a lower bound on the signal to interference plus noise ratio (SINR) requirements. Kaufman and Aazhang [13] proposed that D2D users determined their path loss to the BS according to the received power in the downlink, and then adjusted the transmit power so that the interference caused by D2D users to the BS is minimized. Yu *et al.* [14] improved the system performance in terms of throughput by investigating power control, channel assignment and mode selection. Xu *et al.* [15] proposed an innovative

reverse iterative combinatorial auction mechanism to allocate resources to D2D communications underlying downlink cellular networks. The above works have shown that involving D2D communications can improve the overall system performance by proper resource allocation and reasonable management of interference among cellular and D2D pairs. Compared with the related work, our paper aims to solve the problem of D2D resource allocation in HCNs, and there is no doubt that the interference problems are of great complexity. In this paper, we consider a scheme from the view point of game theory to maximize the system sum rate.

Game theory offers a set of mathematical tools to study the complex interactions among interdependent rational players and to predict their choices of strategies [24]. Besides, with many different game methods included, the game theory has attracted considerable attentions. The related researches utilizing the game theory in the field of wireless communication include the analysis of the resource allocation problems, especially the spectrum allocations in the cellular and heterogeneous networks. Wang *et al.* [16] studied the community-aware D2D resource allocation and further proposed a two-step coalition game to implement effective resource allocation underlying cellular networks. Wang *et al.* [17] proposed a cooperative coalition game to cope with the problem that on-board units might not have the ability to complete the download task of the entire large file from the roadside unit when moving at high speed in vehicular ad hoc networks. In order to improve spectrum efficiency, Li *et al.* [18] proposed a coalition formation game to address the problem of uplink resource allocation for multiple cellular users and D2D pairs. Combining both the interference constraints in the physical domain and social connections in the social domain, Zhao *et al.* [19] proposed a social group utility maximization game based D2D resource allocation scheme to maximize each D2D user's social group utility. However, the coalition game in related work aims to find a coalitional structure that maximizes the individual payoffs of the players, while we entail finding a structure that maximizes the total utility.

mm-wave communication is considered to be one of the most concerned candidate technologies for the fifth generation (5G). The fact that lower frequencies of the radio spectrum have become saturated and are unable to meet the exponential growth in traffic demand, has motivated the exploration of the under-utilized mm-wave frequency spectrum for future high-speed broadband cellular networks [21]–[23]. However, mm-wave communications have unique characteristics that are different from traditional cellular networks. On the one hand, mm-wave communication is typically characterized by transmission and reception with very narrow beams and highly directional antenna. On the other hand, mm-wave communication suffers a much larger propagation loss due to the high carrier frequency, and mm-wave links are easily blocked by human body and other obstacles. Consequently, network congestion may happen in mm-wave networks. There are some works on utilizing mm-wave band in wireless network. Ai *et al.* [6] performed some measurements and simulations on indoor mm-wave massive multiple-input multiple-output (MIMO) channel at a band in 26 GHz. Shariat *et al.* [25]

presented some important findings in designing radio resource management (RRM) functionalities of mm-wave in conjunction with heterogeneous network in both backhaul and access links. Rebato *et al.* [26] proposed an effective novel hybrid spectrum access scheme consisted of the exclusive low frequency carrier and the pooled high frequency carrier for mm-wave networks. Niu *et al.* [5] developed an energy-efficient mm-wave backhauling scheme to deal with the joint optimization problem of concurrent transmission scheduling and power control of small cells densely deployed in HCNs.

III. SYSTEM OVERVIEW AND PROBLEM FORMULATION

In this section, we first give a system overview for D2D communications underlying HCN, and then formulate the resource allocation problem by defining optimization utility function that reflects the system performance in terms of system sum rate.

A. System Description

We consider a scenario of a single cell coupled with all the users under its coverage. In our investigated system, we focus on the intra-cell interference generated by the users sharing the same frequency band. Since the heterogeneous network consists of the cellular band and mm-wave band, there are two kinds of modes to select for each D2D pair. One is to share the uplink spectrum resource of one cellular user, and the other is to use the resource in mm-wave band. On the one hand, we consider the cellular D2D network, where the BS is equipped with omnidirectional antennas for cellular communications. We assume that the cellular users share their uplink resources with D2D communications when the cellular access mode is selected by D2D pairs, and one cellular user's spectrum resource can be shared with multiple D2D pairs to achieve the maximum spectral efficiency, while we also assume that a D2D pair shares no more than one cellular user's uplink resource for the purpose of reducing interference caused by D2D communications and decreasing the corresponding complexity. In addition, it is supposed that the subcarrier channels occupied by cellular users are mutually independent for analytical tractability. In other words, D2D pairs will not interfere with each other when sharing different cellular users' uplink spectrum resources in cellular D2D network. On the other hand, we consider the mm-wave D2D network, which doesn't require infrastructure such as BSs. Millimeter wave communication is equipped with the highly directional antenna in order to achieve the directional transmission and reception between D2D users in mm-wave band [5]. With highly directional antenna arrays in mm-wave, D2D pairs are able to share the same radio resource. As illustrated in Fig. 2, there exists two cellular users c_1 and c_2 , and the D2D pair (d_1^1, d_1^2) occupies the spectrum resource of c_1 , while D2D pairs (d_2^1, d_2^2) and (d_3^1, d_3^2) occupy the spectrum resource of c_2 . Besides, D2D pairs (d_4^1, d_4^2) and (d_5^1, d_5^2) use the spectrum resource in the mm-wave band. On the whole, we only need to focus on the analysis of the signal interference between D2D pairs in mm-wave band and the signal interference among cellular users and D2D pairs in cellular band.

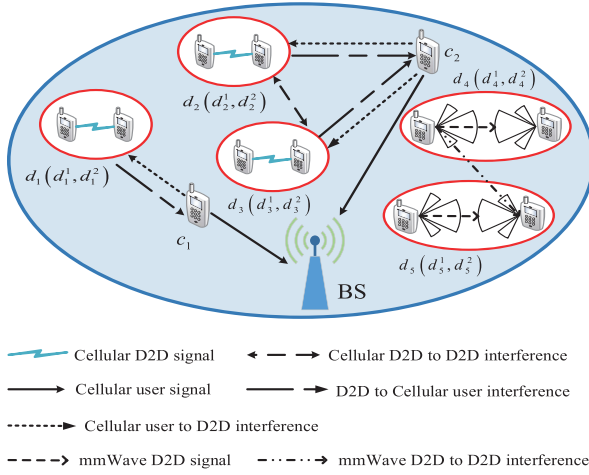


Fig. 2. Illustration of the resource sharing of D2D communications underlaying HCN, where there are 2 cellular users, c_1 and c_2 , and 5 D2D pairs.

In such a system, we concentrate on assigning appropriate uplink spectrum resources occupied by the cellular users or mm-wave radio resource to D2D pairs in order to enhance the whole network performance. Since the D2D pair shares the same spectrum resources with the cellular users or with other D2D pairs in mm-wave band, as the result of that, the system performance will be reduced to compensate the interference. In order to maximize the system performance, what we should do is to properly manage the interference and limit the interference as much as possible. As shown in Fig. 2, there are three kinds of interference in cellular D2D network, such as cellular D2D to D2D interference, D2D to cellular user interference and cellular user to D2D interference. The cellular user and its corresponding D2D pairs interfere with each other because they share the same uplink spectrum resources. The received signals at the BS from the cellular user c are interfered by the transmitters of D2D pairs sharing the same spectrum resource of c . The signal at the D2D receiver d is interfered by the cellular user c and other D2D links sharing the same spectrum resource of c . On the other hand, there exists just one kind of interference in mm-wave D2D network and the D2D pairs are mutually interfered as they use the same spectrum resource in mm-wave band.

B. System Model

In the system, we assume there are C cellular users labeled as the set of $\mathbf{C} = \{c_1, c_2, \dots, c_C\}$ that share their uplink resources with D2D pairs. Moreover, we denote the set of D pairs of D2D users by \mathbf{D} , written as $\mathbf{D} = \{d_1, d_2, \dots, d_D\}$. Every D2D pair independently randomly chooses to share the resource of any cellular user $c_i, \forall c_i \in \mathbf{C}$ or the resource in mm-wave band. To better reflect the spectrum resource usage relationship, we define a binary variable a_d for each D2D pair d to represent whether the cellular or mm-wave frequency band is selected. If the cellular frequency band is selected, $a_d = 1$; otherwise, $a_d = 0$. Besides, we define another binary variable $x_{c,d}$ to indicate whether the uplink spectrum resource of cellular user c is

shared by $d, \forall c \in \mathbf{C}, d \in \mathbf{D}$, where if $x_{c,d} = 1$, it means that the resource blocks of cellular user c are allocated to the D2D pair d , otherwise, $x_{c,d} = 0$. We analyze the constraints of $x_{c,d}$. First, each D2D pair can share the uplink spectrum resource from no more than one cellular user, which can be expressed as $\sum_{c \in \mathbf{C}} x_{c,d} \leq 1, \forall d \in \mathbf{D}$. Second, $x_{c,d}$ is equal to a_d for all D2D pairs, which can be expressed as $\sum_{c \in \mathbf{C}} x_{c,d} = a_d$. On the one hand, sharing the spectrum resource of one cellular user by multiple D2D pairs is allowed in this sharing model in order to increase the spectrum resource reuse ratio. On the other hand, it is also possible for D2D pairs to occur on the same part of the mm-wave spectrum resource.

To maximize the network performance in terms of system sum rate, we should consider the key part of SINR. Assuming that in the cellular D2D network, we adopt the channel model of Rayleigh for small-scale fading with the propagation loss factor n , under which the instantaneous channel taps are the function of time and spatial locations [27]. The power or second-order statistic of the channel, denoted by $|h_0|^2$, is a constant within the BS's coverage area. For communication link i , we denote its sender and receiver by s_i and r_i , respectively. According to the path loss model, we derive the expression of the received power at r_i from s_i as $P_r^c(i, i) = |h_0|^2 \cdot G_t \cdot G_r \cdot l_{ii}^{-n} \cdot P_c$, where P_c is the cellular transmission power, l_{ii} is the distance between s_i and r_i , n is the path-loss exponent, h_0 is a complex Gaussian random variable with unit variance and zero mean, G_t is the transmit antenna gain and G_r is the receive antenna gain. Both of them are constants. The received SINR at r_i from s_i can be expressed as

$$\text{SINR}_i^c = \frac{|h_0|^2 G_t G_r l_{ii}^{-n} P_c}{P_{\text{int},i}^c + N_{0c} W_c}, \quad (1)$$

where $P_{\text{int},i}^c$ is the interference signal power received by user r_i , N_{0c} is the cellular onesided power spectral density of white Gaussian noise, and W_c is the cellular subcarrier bandwidth.

Similarly, we assume that in the mm-wave D2D network, the received power at r_i from s_i can be calculated as

$$P_r^m(i, i) = k_0 G_t(i, i) G_r(i, i) l_{ii}^{-n} P_m. \quad (2)$$

For two mutually independent communication links i and j , the received interference at r_i from s_j can be calculated as

$$P_r^m(j, i) = \rho k_0 G_t(j, i) G_r(j, i) l_{ji}^{-n} P_m, \quad (3)$$

where k_0 is a constant coefficient and proportional to $(\frac{\lambda}{4\pi})^2$ (λ denotes the wavelength), ρ denotes the multi-user interference (MUI) factor related to the cross correlation of signals from different links, and P_m is the transmitted power of mm-wave [28]. Unlike the assumption in cellular D2D network, the antenna gain of s_i in the direction of $s_i \rightarrow r_i$ is denoted by $G_t(i, i)$ and is no longer a constant. The antenna gain of r_i in the direction of $s_i \rightarrow r_i$ is denoted by $G_r(i, i)$. Thus, the received SINR at r_i can be expressed as

$$\text{SINR}_i^m = \frac{P_r^m(i, i)}{P_{\text{int},i}^m + N_{0m} W_m}, \quad (4)$$

where $P_{int,i}^m$ is the interference signal power received by user r_i , N_{0m} is the mm-wave onesided power spectral density of white Gaussian noise, and W_m is the bandwidth of mm-wave communication.

In the case of cellular communication, we abbreviate the transmit and receive antenna gain of device and BS as G_0 and G_b , respectively, since they are taken the fixed value in cellular D2D network. Then, we are able to obtain the uplink transmission rate corresponding to cellular users and D2D pairs. The BS receiving signal from the cellular user subjects to interference from D2D pairs referred to that occupying the same spectrum resource with cellular user. Therefore, the interference power at the BS for cellular user c can be expressed as

$$P_{int,c} = \sum_{d \in \mathbf{D}} x_{c,d} |h_0|^2 G_0 G_b l_{db}^{-n} P_c. \quad (5)$$

According to Shannon's channel capacity, the uplink channel rate of the cellular user c , denoted by R_c , is

$$R_c = W_c \log_2 \left(1 + \frac{|h_0|^2 G_0 G_b l_{cb}^{-n} P_c}{\sum_{d \in \mathbf{D}} x_{c,d} |h_0|^2 G_0 G_b l_{db}^{-n} P_c + N_{0c} W_c} \right). \quad (6)$$

The D2D receiver d suffers interference from the cellular user c and the other D2D pairs sharing the same spectrum resource of c . Therefore, we can get the following expression of interference power for D2D receiver d , denoted by $P_{int,d}^c$.

$$P_{int,d}^c = \sum_{c \in \mathbf{C}} x_{c,d} |h_0|^2 G_0^2 l_{cd}^{-n} P_c + \sum_{d' \in \mathbf{D} \setminus \{d\}} \sum_{c \in \mathbf{C}} x_{c,d} x_{c,d'} |h_0|^2 G_0^2 l_{d'd}^{-n} P_c. \quad (7)$$

According to (7), we can obtain the received SINR at the D2D receiver d , denoted by $SINR_d^c$, as follows.

$$SINR_d^c = \frac{|h_0|^2 G_0^2 l_{dd}^{-n} P_c}{P_{int,d}^c + N_{0c} W_c}. \quad (8)$$

In the case of mm-wave communication, we can derive the transmission rate of D2D pairs similarly. The interference of D2D receiver d is from the other D2D pairs in mm-wave band. Thus, we can obtain the interference power from the other D2D pairs for D2D receiver d , denoted by $P_{int,d}^m$, as follows.

$$P_{int,d}^m = \sum_{d' \in \mathbf{D} \setminus \{d\}} (1 - a_{d'}) \rho k_0 G_t(d', d) G_r(d', d) l_{d'd}^{-n} P_m. \quad (9)$$

According to (9), we can get the following received SINR at the D2D receiver d , denoted by $SINR_d^m$.

$$SINR_d^m = \frac{k_0 G_t(d, d) G_r(d, d) l_{dd}^{-n} P_m}{P_{int,d}^m + N_{0m} W_m}. \quad (10)$$

Combining the $SINR_d^c$ in cellular D2D network and the $SINR_d^m$ in mm-wave D2D network, the SINR received by D2D receiver d in HCN, denoted by $SINR_d$, can be calculated as

$$SINR_d = a_d SINR_d^c + (1 - a_d) SINR_d^m. \quad (11)$$

The achievable channel rate for the D2D pair d , denoted by R_d , is give in (12), shown at the top of the next page.

Thus, the achieved system sum rate considering all the cellular users and D2D pairs in HCN, denoted by R , can be obtained as

$$R = \sum_{c \in \mathbf{C}} R_c + \sum_{d \in \mathbf{D}} (a_d R_d + (1 - a_d)(1 - P_{out:d,d}) R_d), \quad (13)$$

where $P_{out:d,d}$ denotes the probability of blockage in the LOS path between the sender and the receiver of D2D pair d in mm-wave band. It can be expressed as $P_{out:i,j} = 1 - e^{-\beta l_{ij}}$, where l_{ij} is the distance between users i and j , and β is the parameter used to reflect the density and size of obstacles, which result in an interruption caused by blockage [29].

C. Problem Formulation

Obviously, the system sum rate is related to the resource sharing relations $x_{c,d}$ and $a_d, \forall c \in \mathbf{C}, d \in \mathbf{D}$. In view of the relationship between these two binary variables, $\sum_{c \in \mathbf{C}} x_{c,d} = a_d, \forall d \in \mathbf{D}$, we can define a system utility function that reflects the network performance as the system sum rate, denoted by $R(\mathbf{X})$, where \mathbf{X} is the matrix of $x_{c,d}, \forall c \in \mathbf{C}, d \in \mathbf{D}$. Therefore, based on the above analysis, the problem of determining the optimal resource allocation strategy in the D2D communications underlaying HCN to maximize the system sum rate can be formulated as follows.

$$\begin{aligned} \max \quad & R(\mathbf{X}) \\ \text{s.t.} \quad & \begin{cases} x_{c,d} \in \{0, 1\}, \quad \forall d \in \mathbf{D}, c \in \mathbf{C}; \\ \sum_{c \in \mathbf{C}} x_{c,d} \leq 1, \quad \forall d \in \mathbf{D} \end{cases} \end{aligned} \quad (14)$$

This is a nonlinear integer programming problem, where $x_{c,d}$ is the integer binary variable. In the formulated problem, the optimization utility function in (14) has no obvious increasing or concave properties with $x_{c,d}$ even the constraint is linear. Obviously, this problem is NP-complete and it is more complex compared with the 0-1 Knapsack problem [30]. Our optimization problem aims to maximize the system sum rate. In the next section, we propose a coalition formation algorithm from the perspective of game theory to solve the problem with low complexity. For each D2D pair d in the system, or equivalently each player in the game, it makes a decision on selecting the mm-wave band or sharing the spectrum of the cellular user c ($c \in \mathbf{C}$), only for making a greater contribution to the system utility function.

IV. COALITIONAL GAME APPROACH

In this section, we present the coalition game from the view point of game theory to solve the formulated resource sharing problem. Based on it, the coalition formation algorithm is proposed.

A. Coalitional Game Formulation

The formulated optimization problem aims to maximize the overall system performance. Based on the problem, we introduce a coalition game theory model, where the D2D pairs

$$\begin{aligned}
R_d &= a_d W_c \log_2 (1 + SINR_d^c) + (1 - a_d) W_m \log_2 (1 + SINR_d^m) \\
&= a_d W_c \log_2 \left(1 + \frac{|h_0|^2 G_0^2 l_{dd}^{-n} P_c}{\sum_{c \in \mathbf{C}} x_{c,d} |h_0|^2 G_0^2 l_{cd}^{-n} P_c + \sum_{d' \in \mathbf{D} \setminus \{d\}} \sum_{c \in \mathbf{C}} x_{c,d} x_{c,d'} |h_0|^2 G_0^2 l_{d'd}^{-n} P_c + N_{0c} W_c} \right) \\
&\quad + (1 - a_d) W_m \log_2 \left(1 + \frac{k_0 G_t(d, d) G_r(d, d) l_{dd}^{-n} P_m}{\sum_{d' \in \mathbf{D} \setminus \{d\}} (1 - a_{d'}) \rho k_0 G_t(d', d) G_r(d', d) l_{d'd}^{-n} P_m + N_{0m} W_m} \right). \tag{12}
\end{aligned}$$

tend to form coalitions so that the system utility will improve. In our investigated system, there are C cellular users and D D2D pairs. The D2D pairs can choose to occupy the spectrum resource of any of the C cellular users or use the resource in mm-wave band. Thus, we suppose that there are $C + 1$ coalitions formed by D2D pairs. We denote the coalitions as $F = \{F_{c_1}, F_{c_2}, \dots, F_{c_C}, F_{c_{C+1}}\}$, where $F_{c_x} \cap F_{c_{x'}} = \emptyset$ for any $x \neq x'$, and $\bigcup_{x=1}^{C+1} F_{c_x} = \mathbf{D}$. The cardinality of F is the number of coalitions. We divide the coalitions into two groups for discussion. The first group is composed of coalitions of $F_c \subset F$ ($c \in \mathbf{C}$) sharing the resource with cellular user $c \in \mathbf{C}$. The achieved uplink transmission rate of cellular user c in this case can be written as

$$R_c = W_c \log_2 \left(1 + \frac{|h_0|^2 G_0 G_b l_{cb}^{-n} P_c}{\sum_{d \in F_c} |h_0|^2 G_0 G_b l_{db}^{-n} P_c + N_{0c} W_c} \right). \tag{15}$$

The uplink transmission rate of D2D pair d ($d \in F_c$) is given in (16), shown at the top of the next page.

Consequently, the rate of the uplink channel shared by cellular user c and D2D pairs $d \in F_c$, denoted by $R(F_c)$, is given by

$$R(F_c) = R_c + \sum_{d \in F_c} R_d. \tag{17}$$

The other group is coalition $F_c \subset F$ ($c = c_{C+1}$) sharing the resource in mm-wave band. The channel rate of D2D pair d ($d \in F_c$) can be written as

$$\begin{aligned}
R_d &= W_m \\
&\log_2 \left(1 + \frac{k_0 G_t(d, d) G_r(d, d) l_{dd}^{-n} P_m}{\sum_{d' \in F_c \setminus \{d\}} \rho k_0 G_t(d', d) G_r(d', d) l_{d'd}^{-n} P_m + N_{0m} W_m} \right). \tag{18}
\end{aligned}$$

Therefore, the rate of the channel occupied by D2D pairs $d \in F_c$, denoted by $R(F_c)$, is given by

$$R(F_c) = \sum_{d \in F_c} (1 - P_{out:d,d}) R_d. \tag{19}$$

Obviously the larger the number of D2D pairs in a coalition, the greater the resulting interference among users. In the proposed coalitional game, if all the D2D pairs form a grand coalition to share one cellular user's uplink spectrum resource or the resource in mm-wave band, no D2D pair can

make a greater contribution to the system utility due to the severe interference. Therefore, all the D2D pairs are with little incentive to form a grand coalition. In addition, the mm-wave communication rate is about six orders of magnitude larger than that of cellular communication. Thus, multiple D2D pairs will choose to share the resource in mm-wave band, and some of the coalitions sharing the resources of cellular users may be empty for the purpose of maximizing the system sum rate. In this paper, the D2D resource allocation underlying HCN is modeled in the coalitional game with transferable utilities, where the D2D pairs, as the game players, tend to form coalitions to share the resources of cellular users or mm-wave radio resource in order to maximize the system sum profits. Finally, we define the proposed coalitional game with the transferable utility as follows.

Definition 1: Coalitional Game With Transferable Utility: The concept of coalitional game with transferable utility has been first proposed by von Neumann and Morgenstern [31]. A coalitional game with a transferable utility for D2D resource allocation underlying HCN is defined by a pair (\mathbf{D}, R) , where \mathbf{D} is the set of game players and R is the payoff function. Both of them are the basic elements of game theory. $\forall F_c \subset F$, $R(F_c)$ is a real number, which represents the sum profits contributed by the entire coalition F_c , and it can be assigned to the members of coalition F_c in any random way. Next, we define the coalition game for the proposed resource sharing relations.

Definition 2: Coalitional Game for D2D Resource Allocation: The coalitional game with transferable utility for resource allocation of D2D communications is defined by the triple (\mathbf{D}, R, F) , where the set of the D2D pairs \mathbf{D} is players, R is the transferable utility including the transmission rates of all the users in the coalition, and F is the coalition partition, which can be denoted as $F = \{F_{c_1}, F_{c_2}, \dots, F_{c_C}, F_{c_{C+1}}\}$, where $F_{c_x} \cap F_{c_{x'}} = \emptyset$ for any $x \neq x'$, and $\bigcup_{x=1}^{C+1} F_{c_x} = \mathbf{D}$. It is a strategy for each D2D pair d to make a decision on which coalition to share resources based on the system sum utility.

B. Coalition Formation Algorithm

In this subsection, we devise a coalition formation algorithm for the proposed coalition formation game.

One key point in coalition formation is about what strategy to adopt by each D2D pair. In other words, each D2D pair chooses to join one of the coalitions, and then is able to

$$R_d = W_c \log_2 \left(1 + \frac{|h_0|^2 G_0^2 l_{dd}^{-n} P_c}{|h_0|^2 G_0^2 l_{cd}^{-n} P_c + \sum_{d' \in F_c \setminus \{d\}} |h_0|^2 G_0^2 l_{d'd}^{-n} P_c + N_{0c} W_c} \right). \quad (16)$$

compare and order its potential coalitions based on well-defined preferences. In order to evaluate these preferences, we introduce the concept of preference relation or order in detail [18], [19].

Definition 3: Preference Order

For any D2D pair $i \in \mathbf{D}$, the preference relation or order \succ_i is defined as a complete, reflexive, and transitive binary relation over the set of all coalitions that D2D pair i can possibly form.

Hence, the D2D pairs in our coalitional game have the right to choose to join or leave a coalition according to their preference order, that is to say, the D2D pair tends to join a coalition based on which it prefers to being a member. For any given D2D pair $i \in \mathbf{D}$, $F_c \succ_i F_{c'}$ implies that D2D pair i is more willing to be a member of the coalition $F_c \subset \mathbf{D}$ with $i \in F_c$ than $F_{c'} \subset \mathbf{D}$ with $i \in F_{c'}$, which does not include the case that D2D pair i prefers these two coalitions equally. In different applications, the preferences for D2D pairs can be quantified into different inequalities. In this paper, for any D2D pair $i \in \mathbf{D}$ and $i \in F_c, F_{c'}$, we propose the following preference, which is called the utilitarian order [32].

$$F_c \succ_i F_{c'} \iff R(F_c) + R(F_{c'} \setminus i) > R(F_c \setminus i) + R(F_{c'}). \quad (20)$$

This definition means D2D pair i prefers being a member of coalition F_c than $F_{c'}$ under the condition that the system sum profit increases. For forming coalitions based on the above preference order, we define the switch operation as follows.

Definition 4: Switch Operation: Given a partition $F = \{F_{c_1}, F_{c_2}, \dots, F_{c_C}, F_{c_{C+1}}\}$ of the D2D pairs set \mathbf{D} , if D2D pair $i \in \mathbf{D}$ performs a switch operation from F_c to $F_{c'}$, $F_c \neq F_{c'}$, then the current partition F is modified into a new partition F' such that $F' = (F \setminus \{F_c, F_{c'}\}) \cup \{F_c \setminus \{i\}, F_{c'} \cup \{i\}\}$.

We initialize the system by any random coalition partition $F = \{F_{c_1}, F_{c_2}, \dots, F_{c_C}, F_{c_{C+1}}\}$. For any D2D pair $i \in \mathbf{D}$, we suppose its current coalition is F_c , where $F_c \subset F$. Then, we uniformly randomly choose another coalition $F_{c'}$ and suppose the preference relation $F_{c'} \succ_i F_c$ is satisfied, where $F_{c'} \subset F, F_c \neq F_{c'}$, which means a switch operation from F_c to $F_{c'}$ and the current coalition partition will be updated to a new partition F' as shown in definition 4. Actually, the switch operation can be performed if and only if the preference relation defined in (20) is satisfied. In this mechanism, every D2D pair $i \in \mathbf{D}$ can leave its current coalition and join another coalition, given that the new coalition is strictly preferred through the definition in (20) and the D2D pair can make a greater contribution to the entire system performance in terms of sum rate in the new coalition. In general, our proposed coalition formation game entails finding a coalitional structure that maximizes the total utility rather than the individual payoffs of the players.

Algorithm 1 The Coalition Formation Algorithm for the D2D Pairs Resource Allocation

- 1: Given any partition F_{ini} of the D2D pairs set \mathbf{D} ;
- 2: Set the current partition as $F_{ini} \rightarrow F_{cur}, num = 0$;
- 3: **repeat**
- 4: Choose one D2D pair $i \in \mathbf{D}$ in a pre-determined order, and denote its coalition as $F_c \subset F_{cur}$;
- 5: Uniformly randomly search for another possible coalition $F_{c'} \subset F_{cur}, F_{c'} \neq F_c$;
- 6: Calculate $R(F_c)$ and $R(F_{c'})$;
- 7: **if** The switch operation from F_c to $F_{c'}$ satisfying $F_{c'} \succ_i F_c$ **then**
- 8: $num = 0$;
- 9: D2D pair i leaves its current coalition F_c , and joins the new coalition $F_{c'}$;
- 10: Update the current partition set as follows $(F_{cur} \setminus \{F_{c'}, F_c\}) \cup \{F_c \setminus \{i\}, F_{c'} \cup \{i\}\} \rightarrow F_{cur}$;
- 11: **else**
- 12: $num = num + 1$;
- 13: **end if**
- 14: **until** The partition converges to the final Nash-stable partition F_{fin} .

The coalition formation game is summarized in Algorithm 1, where the D2D pairs make switch operation in a random order. In the algorithm, we first give any partition F_{ini} of the D2D pairs set \mathbf{D} . Then, the system will choose one of the D2D pairs in a pre-determined order in step 4. The selected D2D pair saves the coalition F_c currently located and then uniformly randomly selects another possible coalition $F_{c'}$ in step 5. In step 6, the D2D pair obtains the channel information of both coalitions F_c and $F_{c'}$ from BS. Then, it calculates respectively the received sum rate of these two coalitions and makes a decision on whether to perform the switch operation. If the preference relation is satisfied, we update the current coalition partition and reset the number of consecutive unsuccessful switch operations num to zero. Otherwise, we increase the number of consecutive unsuccessful switch operations by 1. When the value of num is equal to multiply the number of D2D pairs by 10 [19], the algorithm stops iterating and performs operations outside the loop. Finally, the system partition will converge to the final Nash-stable partition F_{fin} after a limited number of switching.

V. THEORETICAL ANALYSIS

A. Convergence

In this subsection, the convergence of the proposed coalition formation algorithm is guaranteed as follows [17].

Theorem 1: Starting from any initial coalitional structure F_{ini} , the proposed coalition formation algorithm will always converge to a final network partition F_{fin} , which is consisted by a number of disjoint coalitions, after a sequence of switch operations.

Proof: Through careful inspection of the preference defined in (20), we find that each switch operation in Algorithm 1 will either yield an unvisited partition through adopting new strategy or switch existing partitions. As a result, part of coalitions may degenerate into the sets of very few D2D pairs, and even be emptied. The system will form at most $C + 1$ partitions as there is only C cellular users plus one mm-wave band. As the number of partitions for the already given D2D pairs set \mathbf{D} is the Bell number [32], we draw the conclusion that the sequence of switch operations will always terminate and converge to a final partition F_{fin} , which completes the proof that our proposed coalition formation algorithm is convergent. ■

B. Stability

In this subsection, we study the stability of the proposed coalition formation algorithm by using the definition from the hedonic games as follows [33].

Definition 5: Nash-stable Structure: A coalitional partition $F = \{F_{c1}, F_{c2}, \dots, F_{cc}, F_{cc+1}\}$ is Nash-stable, if $\forall i \in \mathbf{D}, i \in F_c \subset F, F_c \succ_i F_{c'} \cup \{i\}$ for all $F_{c'} \subset F, F_{c'} \neq F_c$.

Theorem 2: The final partition F_{fin} in our coalition formation algorithm is Nash-stable.

Proof: The coalition game has the Nash-stable coalitional structure if no D2D pair can make its contribution to the entire system increased by changing its resource sharing strategy. $F_c^* = \arg \max_{F_c} R(F), \forall F_c \subset F$, and $F^* = \{F_{c1}^*, F_{c2}^*, \dots, F_{cc}^*, F_{cc+1}^*\}$ is the final Nash-stable coalitional structure. We prove the stability by contradiction. Assuming that the final formed coalition partition F_{fin} is not Nash-stable. In other words, there exists a D2D pair $i \in \mathbf{D}$, and its located coalition currently and randomly selected new coalition are denoted by F_c and $F_{c'}$ respectively. These two coalitions meet the preference relation $F_{c'} \cup \{i\} \succ_i F_c$. Consequently, D2D pair will perform the operations leaving its current coalition F_c and joining the new coalition $F_{c'}$, which means that F_{fin} will be updated and it is not the final partition. Thus, we complete the proof that the final partition F_{fin} of our proposed coalition formation algorithm is Nash-stable. ■

C. Optimality

Theorem 3: The solution obtained by our proposed algorithm corresponds to an optimal system performance.

Proof: The total utility achieved by our proposed coalition formation algorithm is convergent with a sufficiently large number of iterations. In Algorithm 1, we set the termination condition to be that the number of consecutive unsuccessful switch operations num is equal to the product of the number of D2D pairs and 10. On the other hand, our scheme only involves one-step switching and it has the limitation of allowing multiple D2D pairs to perform switch operations

simultaneously. Thus, the solution obtained by Algorithm 1 is near-optimal compared with the solution obtained by the exhaustive search method. From the Fig. 4(a) and Fig. 4(b) in Section VI, the gap between our scheme and the optimal solution is quite small and the performance of our proposed algorithm is guaranteed. Besides, the in-depth analysis of the performance bound will be carried out in the future work. ■

D. Complexity

Theorem 4: Given the total number of iterations N , the computational complexity of Algorithm 1 can be approximated as $O(N)$.

Proof: In each iteration of Algorithm 1, the selected D2D pair calculates the total utility of currently located coalition and another possible coalition, respectively. Then, it makes a decision on whether to perform a switch operation. Thus, there is at most 1 switch operation to be considered in each iteration, and the complexity lies in the number of iterations. From the Fig. 13, we can see the computational complexity of Algorithm 1 is extremely low. ■

VI. PERFORMANCE EVALUATION

In this section, we evaluate the performance of our proposed coalition game under various system parameters. Specially, we compare our scheme with other four schemes in terms of system sum rate. Besides, we give the necessary analysis for the obtained simulation results.

A. Simulation Setup

In the simulation, we consider a single cell scenario, where D2D pairs and cellular users are uniform randomly distributed in a square area of $500m \times 500m$ with the base station in the center. For a fixed number of cellular users and D2D pairs, we repeat the simulation by 20 times and then average the results of positions in order to obtain a more reliable location layout. Not only the path-loss model is considered for cellular and D2D links, but also the shadow fading. Besides, we set the path-loss exponent in free space propagation model to be 2. On the one hand, when two D2D users are physically in close proximity, the D2D communication channel is established. In our simulation, we provide an upper bound on the distance between two D2D users. On the other hand, the widely used realistic directional antenna model is adopted in mm-wave D2D network, which is a main lobe of Gaussian form in linear scale and constant level of side lobes [5]. Based on this model, the gain of a directional antenna in units of decibel (dB), denoted by $G(\theta)$, can be expressed as

$$G(\theta) = \begin{cases} G_0 - 3.01 \cdot \left(\frac{2\theta}{\theta_{-3dB}} \right)^2, & 0^\circ \leq \theta \leq \theta_{ml}/2; \\ G_{sl}, & \theta_{ml}/2 \leq \theta \leq 180^\circ; \end{cases} \quad (21)$$

where θ denotes an arbitrary angle within the range $[0^\circ, 180^\circ]$, θ_{-3dB} denotes the angle of the half-power beam width, and θ_{ml} denotes the main lobe width in units of degrees. The relationship between θ_{ml} and θ_{-3dB} is $\theta_{ml} = 2.6 \cdot \theta_{-3dB}$. G_0 is the

TABLE I
SIMULATION PARAMETERS

| Parameter | Symbol | Value |
|---------------------------------|-----------------|----------------|
| mm-wave bandwidth | W_m | 2160 MHz |
| Cellular carrier bandwidth | W_c | 15 KHz |
| mm-wave noise spectral density | N_{0m} | -134 dBm/MHz |
| Cellular noise spectral density | N_{0c} | -174 dBm/Hz |
| mm-wave transmission power | P_m | 20 dBm |
| Cellular transmission power | P_c | 23 dBm |
| Path loss exponent | n | 2 |
| MUI factor | ρ | 1 |
| Half-power beamwidth | θ_{-3dB} | 30° |
| Blockage parameter | β | 0.01 |
| Antenna gains of device | G_0 | 0.5 dBi |
| Antenna gains of BS | G_b | 14 dBi |
| Maximum distance of D2D | r | $10\sqrt{2}$ m |

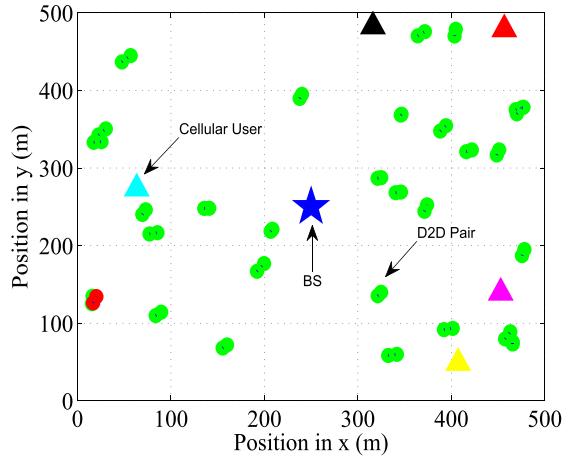


Fig. 3. A snapshot of a final coalition structure resulting from CG for a network of 5 cellular users and 30 D2D pairs.

maximum antenna gain, and can be expressed as

$$G_0 = 10 \log \left(\frac{1.6162}{\sin(\theta_{-3dB}/2)} \right)^2. \quad (22)$$

G_{sl} denotes the side lobe gain, which can be obtained by

$$G_{sl} = -0.4111 \cdot \ln(\theta_{-3dB}) - 10.579. \quad (23)$$

The simulation parameters are summarized in Table I [5]. In order to illustrate how cellular and D2D users are distributed and how to share resources, we plot the positions of the base station, cellular users and D2D pairs together in an instance by randomly generating a network consisting of 5 cellular users and 30 D2D pairs in Fig. 3. Besides, we show a snapshot of a final coalition structure resulting from our coalitional formation algorithm. In the figure, the base station, cellular users and D2D pairs are represented by pentacle, triangle and circle respectively. Five cellular users and thirty D2D pairs form six coalitions, and they are marked by different colors of red, green, cyan, dark, yellow and magenta, respectively.

In order to show the advantage of our proposed coalition game in improving system performance in terms of system sum rate $R(F)$, which includes the communication rates of all cellular users and D2D pairs, we compare our scheme, labeled as **Coalition Game** (CG), with four other schemes:

a) **Full Mm-wave Communication** (FMC), where all the D2D pairs are interconnected via direct D2D communications in mm-wave band, and each cellular user occupies one of the cellular carrier channels without spectrum sharing.

b) **Random Communication** (RC), where the system allocates the communication resources to the D2D pairs in a uniform randomly manner. In other words, for any D2D pair, the system randomly selects a cellular user's spectrum resource or the resource in mm-wave band.

c) **Cellular Coalition Game** (CCG), which utilizes coalition game to cope with the problem of the resource allocation among cellular bands for multiple D2D pairs in cellular network. In order to maximize the system total utility, the algorithm performs switch operations based on well-defined preference order with a limited number of iterations.

d) **Full Cellular Communication** (FCC), which uniform randomly allocates cellular users' uplink spectrum resources to the D2D pairs. Generally speaking, this kind of method is similar to RC, and the difference is that this scheme does not involve mm-wave. Since the transmission rate of cellular communication is much smaller than that of mm-wave communication, this kind of method represents the worst case of the system performance in terms of sum rate compared with above methods.

B. Compared With the Optimal Solution

In this subsection, we compare the performance of CG with the optimal solution, labeled as OS, which is obtained by the traditional exhaustive search method. In view of the highly complexity of this method, we set the number of D2D pairs to be 10 and vary the number of cellular users to be 1 to 8 to obtain the simulation results shown in Fig. 4(a), while set the number of cellular users to be 1 and vary the number of D2D pairs to be 1 to 8 to obtain the simulation results shown in Fig. 4(b). From these two figures, we can see the system sum rate achieved by CG, shown by the dot and dash curve, has an excellent approximation to that achieved by OS, shown by solid line curve. In order to further demonstrate our proposed scheme CG converges close to the OS, we analyze the simulation results in detail and calculate the average deviation between the results obtained by CG and OS, which is expressed as follows.

$$\text{Average Deviation} = \frac{1}{8} \sum_{n=1}^8 \frac{R_{OS}(n) - R_{CG}(n)}{R_{OS}(n)}, \quad (24)$$

where $R_{OS}(n)$ and $R_{CG}(n)$ denote the system sum rate obtained by OS and CG, respectively, with the number of cellular users or D2D pairs n . As a result, the average deviation between the CG and OS is about 0.9% in Fig. 4(a), while the average deviation is about 0.4% in Fig. 4(b). Thus, we complete the demonstration that our proposed coalition game can achieve the system sum rate which is close to the optimal solution of the resource allocation problem.

C. System Sum Rate

In this subsection, we evaluate the performance of our proposed coalition game based resource allocation scheme

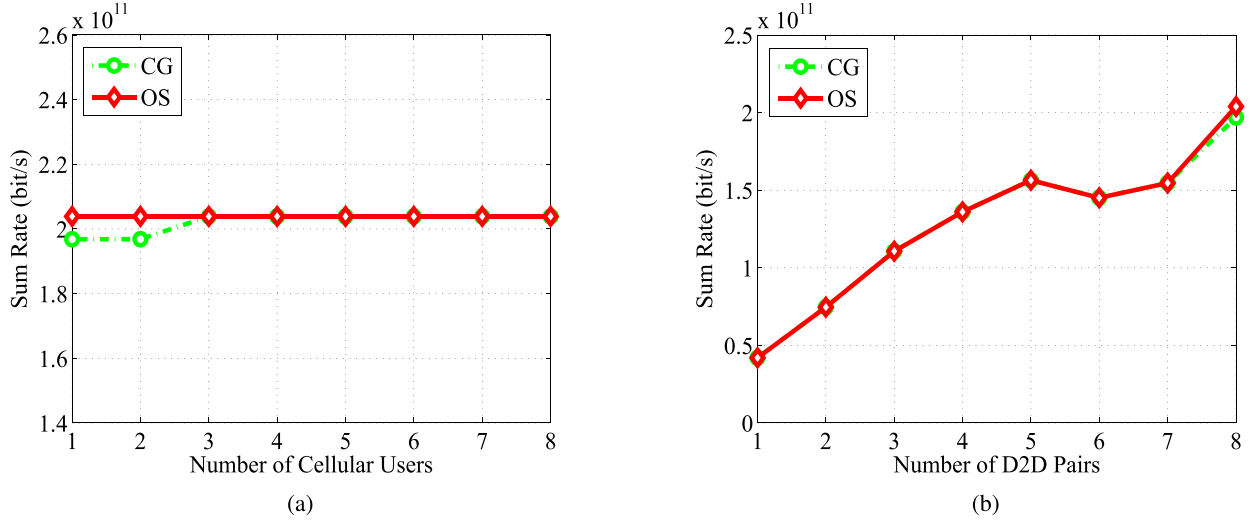


Fig. 4. System sum rate comparison of CG and OS with (a) different number of cellular users and (b) different number of D2D pairs.

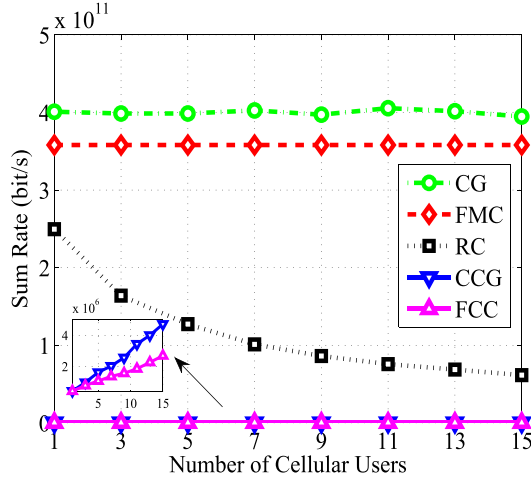


Fig. 5. System sum rate comparison of five resource allocation algorithms with different number of cellular users.

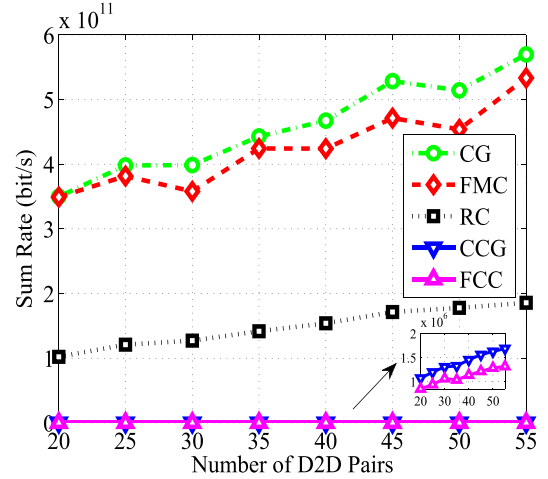


Fig. 6. System sum rate comparison of five resource allocation algorithms with different number of D2D pairs.

under various system parameters, and then demonstrate the advantage of this algorithm compared with four other schemes.

In Fig. 5, we set the number of D2D pairs to be 30 and the other parameter settings are shown in Table I. Then, we plot the system sum rate comparison of five schemes varying the number of cellular users from 1 to 15. From the figure, we can observe that the system sum rate of CG has almost no changes as the number of cellular users increases. It is because that the mm-wave communication rate is much greater than that of cellular communication, which results in the increase in the number of D2D pairs using the spectrum resource in mm-wave band in order to maximize the system sum rate. In other words, the utility contributed by cellular users and D2D pairs in cellular band accounts for a very small proportion of total system utility. At the same time, the randomness of the CG leads to slight fluctuations in the curve. Comparing these five schemes, the system sum rate received by adopting CG is much larger than other schemes. When the number of cellular users is equal to 15, the sum rate of CG is larger

than that of FMC and RC about 10% and 543%, respectively. In addition, with the number of cellular users increased, more D2D pairs will uniform randomly choose to share the spectrum resources with cellular users and the number of D2D pairs using the resource in mm-wave band is decreased, which explains the change in the RC curve. CCG and FCC increase as the number of cellular users increases. The reason is that the bandwidth resource for the D2D transmission increases. Meanwhile, the cellular users still make a contribution to the system sum rate. For CG, FMC and RC, involving mm-wave D2D communications can offload cellular traffic and improve the system performance at the same time.

In Fig. 6, we set the number of cellular users to be 5 and vary the number of D2D pairs to be 20 to 55. From the figure, we can see the proposed CG algorithm performs much better than other schemes. When the number of D2D pairs is equal to 55, the sum rate of CG is larger than that of FMC and RC about 7% and 207%, respectively. Fig. 6 indicates the system sum rate of five schemes increases as the number of D2D pairs

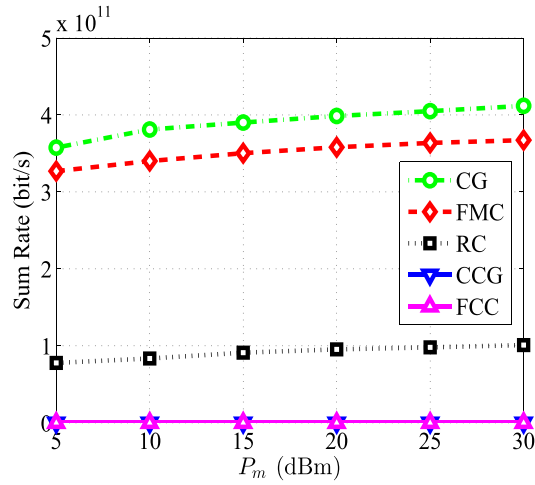


Fig. 7. System sum rate comparison of five resource allocation algorithms with different P_m .

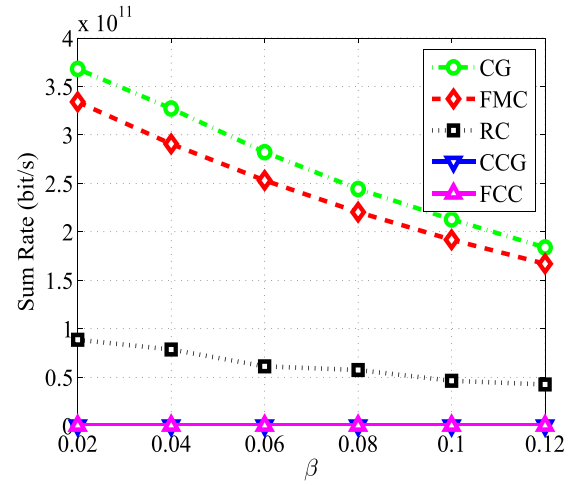


Fig. 9. System sum rate comparison of five resource allocation algorithms with different β .

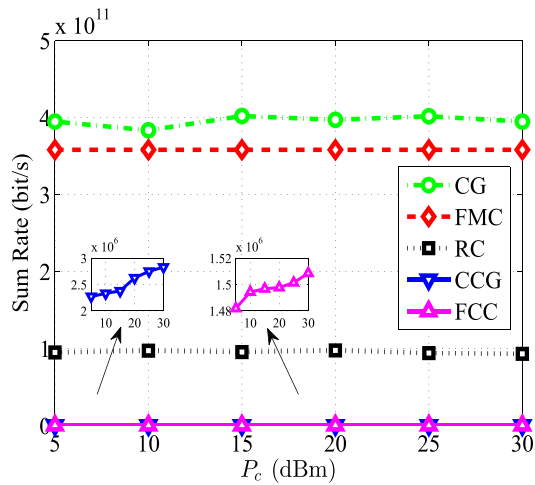


Fig. 8. System sum rate comparison of five resource allocation algorithms with different P_c .

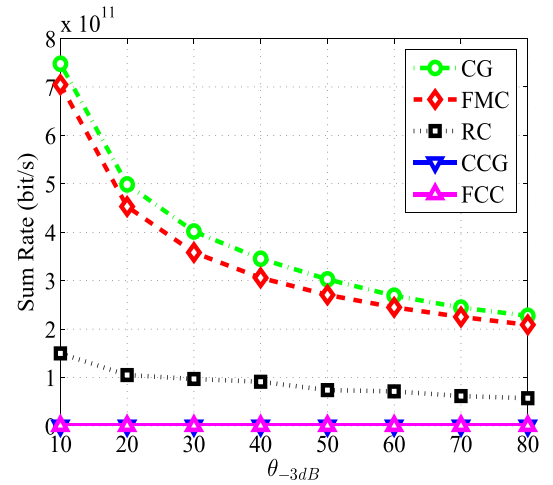


Fig. 10. System sum rate comparison of five resource allocation algorithms with different θ_{-3dB} .

increases. At the same time, different number of D2D pairs makes the change of positions in each simulation, which leads to individual drop points in CG and FMC. With more D2D pairs included in the network, the spectrum utilization can be improved, while the interference caused by spectrum sharing also increases, which constraints the system performance. Besides, the FCC still gets the worst performance, while the FMC, RC and CCG achieve the middle performance.

We set the number of cellular users and D2D pairs to be 8 and 30. Fig. 7 indicates the system sum rate increases with the mm-wave transmission power P_m varied from 5 to 30 dBm in CG, FMC and RC. These three curves grow slowly as the result that the corresponding interference power increases and the improvement in sum rate would be less with the P_m increased. Compared the behaviors of different schemes, we observe that the CG obtains the highest system sum rate. FCC obtains the lowest system sum rate, while FMC, RC and CCG perform medially. When the mm-wave transmission power P_m is equal to 30 dBm, the sum rate of CG is larger than that of FMC and RC about 12% and 307%, respectively.

Similarly, Fig. 8 indicates the system sum rate of CCG and FCC increases with the cellular transmission power P_c varied from 5 to 30 dBm, while the effect of P_c on CG, FMC and RC is not significant. When the cellular transmission power P_c is equal to 30 dBm, the sum rate of CG is larger than that of FMC and RC about 10% and 325%, respectively.

In Fig. 9, we set the number of cellular users and D2D pairs to be 8 and 30, respectively, and then vary the β from 0.02 to 0.12. In terms of the impact of the blockage parameter that captures the density and size of obstacles, we observe that our proposed scheme again has the best performance. When the blockage parameter β is equal to 0.12, the sum rate of CG is larger than that of FMC and RC about 10% and 332%, respectively. The greater β means obstacles with higher density and larger size, which results in higher blockage probability. In other words, the rate of the mm-wave communication channel shared by D2D pairs decreases due to unreliable direct D2D connectivity with β increased, which explains the changes of CG, FMC and RC. Besides, the system

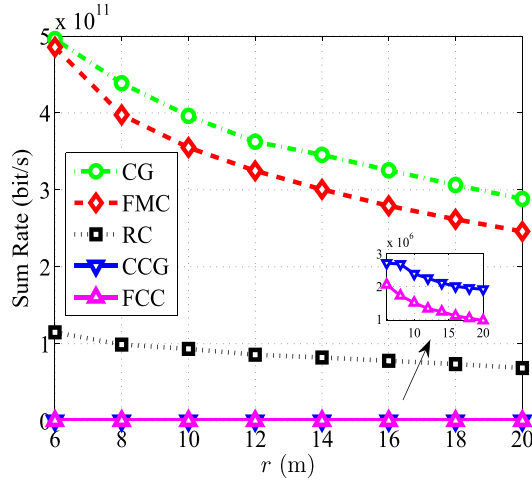


Fig. 11. System sum rate comparison of five resource allocation algorithms with different r .

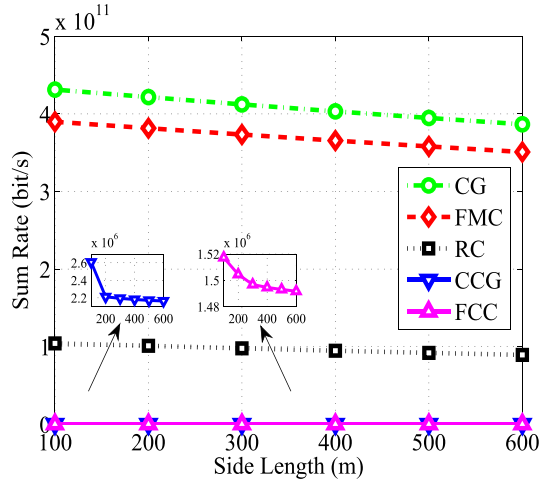


Fig. 12. System sum rate comparison of five resource allocation algorithms with different side length.

sum rate of CCG and FCC keeps at a low level and they are not affected by changing β .

In Fig. 10, we set the number of cellular users and D2D pairs to be 8 and 30, respectively, and then plot the system sum rate comparison of five resource allocation algorithms varying θ_{-3dB} from 10 to 80. The parameter of θ_{-3dB} denotes the angle of the half-power beamwidth adopting the widely used realistic directional antenna model in mm-wave D2D network. As the θ_{-3dB} increases, the system sum rate of CG, FMC and RC decreases. This is because the antenna with larger beamwidth covers the wider area, which causes greater interference toward other D2D pairs in mm-wave band, and furthermore results in the changes in Fig. 10. From the figure, we can see the CG performs better than other schemes. When the θ_{-3dB} is equal to 80, the sum rate of CG is larger than that of FMC and RC about 9% and 298%, respectively.

In Fig. 11, we set the number of cellular users and D2D pairs to be 8 and 30, respectively, and then plot the system sum rate comparison of five resource allocation algorithms varying the maximum distances of both abscissa and ordinate between D2D users from 6 to 20. On the one hand, whether the cellular

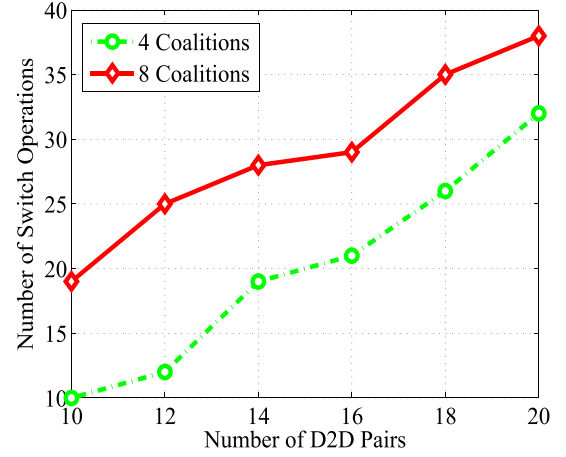


Fig. 13. System convergence rate in terms of the number of switch operations with different number of D2D pairs.

communication mode, or mm-wave communication mode, the increase of r will make the path loss more serious, and thus make the system performance of CG, FMC, RC, CCG and FCC decreased. On the other hand, as the maximum distance of D2D is enlarged, the blockage probability of mm-wave communication link increases and the reliability of the link decreases, which furthermore hampers the performance of CG, FMC and RC.

In Fig. 12, we set the number of cellular users and D2D pairs to be 8 and 30, respectively, and then plot the system sum rate varying the side length of the square from 100 to 600. In order to obtain simulation results under practical scenarios, we modify the maximum distance of D2D as $2\sqrt{2}$, $4\sqrt{2}$, $6\sqrt{2}$, $8\sqrt{2}$, $10\sqrt{2}$ and $12\sqrt{2}$, respectively. As the side length of the square is enlarged, or equivalently the user distribution density is decreased, the path loss of all links and the blockage probability of mm-wave communication links are increased, which directly reduces the system sum rate of all schemes.

D. Convergence Rate

In order to show the convergence rate of our proposed algorithm, we set the number of cellular users to be 3 and 7, or equivalently the number of coalitions to be 4 and 8, and vary the number of D2D pairs to be 10 to 20. In Fig. 13, we show the number of switch operations of CG converging to the final partition. From the figure, we observe that the number of switch operations increases with the number of coalitions or D2D pairs increased. In the cases of 3 and 7 cellular users, the average number of switch operations is from 10 to 32 and 19 to 38, respectively. For the exhaustive search method, each D2D pair can choose to join one of the 8 coalitions when there exists 7 cellular users. The exhaustive search method needs 8^N iterations to find the optimal solution as the number of D2D pairs is set to be N . Therefore, our proposed coalition game algorithm allows D2D pairs and cellular users to form the final Nash-stable partition with extremely fast convergence rate and decreases the computation complexity significantly.

VII. CONCLUSION

In this paper, we investigate the problem of maximizing the system sum rate via resource allocation for D2D communications underlaying HCN combining mm-wave and the traditional cellular band. After formulating the problem of the uplink resource allocation among mm-wave and the cellular band for multiple D2D pairs and cellular users into a non-linear integer programming problem, we propose a coalition game based approach to obtain the near-optimal solution. Through extensive simulations under various system parameters, we demonstrate the superior performance of our proposed coalition game in terms of sum rate compared with four other practical schemes.

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