# Efficient Resource Allocation and Power Control for D2D Communication with RAN Sharing

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Abstract—Device-to-device (D2D) communication improves the performance and flexibility of mobile networks. It enables direct communication between two user equipments (UEs) without the relay via a base station (BS). D2D pairs can share the spectrum resources given to cellular UEs, but doing so causes interference. Thus, how to manage resources and transmitted power for UEs is important. This paper proposes a D2D resource allocation and power control with RAN sharing (DACS) framework, where some service providers (SPs) each possess a part of the BS's resources. Each SP allots its resources to UEs and decides power to satisfy UEs' SINR demands while reducing interference. SPs can borrow resources from each other for serving more UEs. Then, DACS amplifies the power of senders adaptively to better signal quality. Simulation results reveal that DACS can achieve a high service rate, increase energy efficiency, and keep fairness among SPs.

*Index Terms*—D2D communication, energy efficiency, fairness, power control, RAN sharing, resource allocation.

## I. INTRODUCTION

The rapid growth in the number of *user equipments (UEs)* has made spectrum resources scarce. A promising solution is *device-to-device (D2D) communication*. It implements direct communication between two UEs (referred to as a *D2D pair*) without requiring a *base station (BS)* to relay messages. This technique has benefits: 1) The spectral efficiency is improved, as resources are shared by D2D pairs and *cellular UEs (CUs)*. 2) UEs in a D2D pair (called *D2D UEs*, or *DUs* for short) are close to each other, so a D2D sender can reduce its transmitted power (below, we call it *power*) to save energy. 3) Using D2D communication expands a BS's service coverage [1].

In D2D communication, both *resource allocation* and *power control* are critical issues [2]. For a BS, its spectrum resources are divided into *resource blocks* (*RBs*) for allocation [3]. The resource allocation problem asks how to assign the BS's RBs to CUs and D2D pairs. When a CU and D2D pairs share an RB, their senders impose interference. Thus, the power control issue asks how to calculate the power for senders to improve throughput of UEs while mitigating interference.

Considering the massive costs of telecommunications infrastructure, 3GPP proposes *radio access network (RAN)* sharing [4] to help *service providers (SPs)* reduce network deployment and operational expenditure. More concretely, multiple SPs are allowed to share spectrum resources and infrastructure (e.g., BSs). As a result, using RAN sharing can substantially extend services and coverage of a mobile network.

This paper considers D2D communication in a RAN sharing scenario and proposes a D2D resource allocation and power

*control with RAN sharing (DACS)* framework. Each SP first allots its RBs to UEs and finds power to meet SINR demands of UEs while reducing interference. In the case that there are UEs not served, an SP can borrow resources from other SPs for serving them. Our RB lending mechanism takes account of both efficiency and fairness. Then, the power of some senders is moderately adjusted to increase UE throughput. Simulation results demonstrate that DACS can attain a high service rate, raise energy efficiency, and maintain fairness among SPs.

## II. RELATED WORK

The resource allocation problem for D2D communication is NP-hard [5]. Many studies convert it to a matching issue. The work [6] picks a D2D pair to share each CU's RB based on the Hungarian method. Zhou et al. [7] find pairs of CUs and DUs for sharing RBs to raise energy efficiency. The study [8] uses the Gale-Shapley algorithm to assign DUs to reuse RBs given to CUs. However, each RB can be shared by no more than one D2D pair, resulting in low RB utilization.

How to allot resources and power to UEs via vertex coloring is discussed in [9], [10], [11]. They build a graph to express the interference relationship between UEs. Adjacent vertices (i.e., UEs) cannot share RBs. The work [12] treats UEs as players and uses a Stackelberg game for RB allocation. Raghu et al. [13] propose an iterative-based method to manage resources and power. The work [14] applies deep reinforcement learning to D2D resource allocation.

Some studies take RAN sharing into account. The work [15] allots RBs to CUs of different SPs and lets DUs reuse RBs allocated to CUs with the least interference. With binary integer programming, the study [16] formulates the RB allocation problem that considers each SP's cost. In [17], SPs can trade RBs based on channel conditions. The study [18] assigns RBs to UEs by referring to their types to meet QoS requirements. However, these studies assume fixed power for the BS and DU senders (i.e., no power control). The work [19] adjusts power for senders and allows SPs borrowing RBs from each other to increase throughput. However, some SPs may borrow more RBs from others, causing unfairness on SPs' throughput.

## **III. SYSTEM MODEL**

Consider a BS offering a set  $\mathcal{R}$  of downlink RBs. There is also a set  $\mathcal{S}$  of SPs that cooperate on the BS. Each SP  $s_k \in \mathcal{S}$ possesses a subset  $\mathcal{R}_k \subset \mathcal{R}$  of RBs with conditions as follows: all RBs in  $\mathcal{R}$  are assigned to SPs in  $\mathcal{S}$  (i.e.,  $\bigcup_{s_k \in \mathcal{S}} \mathcal{R}_k = \mathcal{R}$ ), each SP holds non-zero RBs (i.e.,  $|\mathcal{R}_k| > 0, \forall s_k \in \mathcal{S}$ ), and no two SPs have overlapped RBs (i.e.,  $\mathcal{R}_k \cap \mathcal{R}_m = \emptyset, s_k \neq s_m$ ). SPs may borrow RBs from each other. This can be done via a *capacity broker*, as proposed by 3GPP, which takes charge of managing requests and leases for resources between SPs [20].

Cellular and D2D communication share the same band, and D2D pairs can reuse RBs assigned to CUs. Each D2D pair has a sender and a receiver (the latter is used to represent the pair). Each DU belongs to only one D2D pair. Let us denote by  $C_k$ and  $\mathcal{D}_k$  the sets of CUs and D2D receivers subscribed to SP  $s_k$ , where  $C_k \cap \mathcal{D}_k = \emptyset$ . Moreover, C and  $\mathcal{D}$  signify all CUs and D2D receivers, that is,  $C = \bigcup_{\forall s_k \in S} C_k$  and  $\mathcal{D} = \bigcup_{\forall s_k \in S} \mathcal{D}_k$ .

RB allocation needs to adhere to five rules. First, two CUs cannot use the same RB. Second, D2D pairs may reuse a CU's RBs. Third, each SP  $s_k$  shall first allot RBs in  $\mathcal{R}_k$  to its UEs. Fourth, after  $s_k$  allots RBs by the 3rd rule,  $s_k$  can lend other SPs RBs in  $\mathcal{R}_k$ . Finally, if a D2D pair's members belong to different SPs, the receiver's SP handles RB allocation.

Suppose that UE  $u_i$  gets data from its sender  $\varepsilon(i)$  on RB  $r_x$ . The strength of  $\varepsilon(i)$ 's signal acquired by  $u_i$  is  $g_{\varepsilon(i),i} \times p_{\varepsilon(i),i}^x$ . Here,  $g_{\varepsilon(i),i}$  and  $p_{\varepsilon(i),i}^x$  denote  $\varepsilon(i)$ 's channel gain (hereinafter called gain) and power on  $r_x$  to send data to  $u_i$ . When UEs  $u_i$  and  $u_j$  share  $r_x$ ,  $u_j$ 's sender  $\varepsilon(j)$  imposes an amount  $\psi_{\varepsilon(j),i}^x$  of interference on  $u_i$ . Then, we can estimate  $u_i$ 's SINR on  $r_x$  as follows:

$$\sigma_{i}^{x} = \frac{z_{i}^{x}(g_{\mathsf{BS},i} \times p_{\mathsf{BS},i}^{x})}{\sum_{u_{j} \in \mathcal{D}} z_{j}^{x} \psi_{\varepsilon(j),i}^{x} + \varphi} \quad \text{if } u_{i} \in \mathcal{C} \text{ (i.e., a CU),} \quad (1)$$
  
$$\sigma_{i}^{x} = \frac{z_{i}^{x}(g_{\varepsilon(i),i} \times p_{\varepsilon(i),i}^{x})}{\sum_{u_{j} \in \mathcal{C}} z_{j}^{x} \psi_{\mathsf{BS},i}^{x} + \sum_{u_{j'} \in \mathcal{D} \setminus \{u_{i}\}} z_{j'}^{x} \psi_{\varepsilon(j'),i}^{x} + \varphi} \quad \text{otherwise (i.e., a D2D receiver).} \quad (2)$$

where  $z_j^x = 1$  if  $u_j$  uses  $r_x$ , or  $z_j^x = 0$  otherwise. In addition,  $\varphi$  indicates the thermal noise.

Let  $v_i^x$  be the indicator to show whether the channel quality of UE  $u_i$  on RB  $r_x$  fulfills its minimal required SINR  $\sigma_i^{\min}$ ; in particular,  $v_i^x = 1$  if  $\sigma_i^x \ge \sigma_i^{\min}$ , or  $v_i^x = 0$  otherwise. Then, our problem can be expressed by

maximize 
$$\sum_{r_x \in \mathcal{R}} \sum_{s_k \in \mathcal{S}} \sum_{u_i \in \mathcal{C}_k \cup \mathcal{D}_k} z_i^x v_i^x$$
, (3)

subject to constraints as follows:

$$\sum_{u_i \in \mathcal{C}} z_i^x \le 1, \sum_{u_i \in \mathcal{C} \cup \mathcal{D}} z_i^x \le \delta, \qquad \forall r_x \in \mathcal{R}, \quad (4)$$

$$z_i^x \in \{0,1\}, v_i^x \in \{0,1\}, \quad \forall r_x \in \mathcal{R}, \forall u_i \in \mathcal{C} \cup \mathcal{D},$$
(5)

$$p_i^{\min} \le p_{\varepsilon(i),i}^x \le p_i^{\max}, \qquad \forall r_x \in \mathcal{R}, \forall u_i \in \mathcal{C} \cup \mathcal{D}.$$
(6)

Here, Eq. (3) is to maximize CUs and D2D pairs served, where a UE  $u_i$  is *served* if it obtains RBs that satisfy SINR demand  $\sigma_i^{\min}$ . Then, Eq. (4) indicates that two CUs cannot employ the same RB and gives the maximum number of UEs in  $\mathcal{C} \cup \mathcal{D}$ allowed to share each RB ( $\delta \in \mathbb{Z}^+$ ). In Eq. (5),  $z_i^x$  and  $v_i^x$  are indicators whose values are 0 or 1. Eq. (6) imposes lower and upper bounds (i.e.,  $p_i^{\min}$  and  $p_i^{\max}$ ) on the power of  $u_i$ 's sender.

Algorithm 1: Basic Resource Allocation (BRA)

1 foreach  $u_i \in \mathcal{C} \cup \mathcal{D}$  do foreach  $u_i \in (\mathcal{C} \cup \mathcal{D}) \setminus \{u_i\}$  do 2 if  $g_{\varepsilon(i),j} > g_{th}$  then 3  $\zeta_{i,j} \leftarrow 1;$ 4 5 foreach  $s_k \in \mathcal{S}$  do  $\mathcal{C}'_k \leftarrow f_{\text{SRT}}^{\text{DEC}}(\mathcal{C}_k, g_{\text{BS},i}), \ \mathcal{D}'_k \leftarrow f_{\text{SRT}}^{\text{DEC}}(\mathcal{D}_k, g_{\text{BS},i});$ 6 foreach  $u_i \in \mathcal{C}'_k$  do 7 foreach  $r_x \in \mathcal{R}_k$  do 8 if  $\mathcal{Q}_x \cap \mathcal{C} = \emptyset$  and  $\zeta_{i,j} = 0$ ,  $\forall u_j \in \mathcal{Q}_x$  then  $\sqcup$  Move  $u_i$  from  $\mathcal{C}'_k$  to  $\mathcal{Q}_x$  and break; 9 10 foreach  $u_i \in \mathcal{D}'_k$  do 11 foreach  $r_x \in \mathcal{R}_k$  do 12 if  $|\mathcal{Q}_x| < \delta$  and  $\zeta_{i,j} = 0, \forall u_j \in \mathcal{Q}_x$  then 13 Move  $u_i$  from  $\mathcal{D}'_k$  to  $\mathcal{Q}_x$  and break; 14 foreach  $r_x \in \mathcal{R}_k$  do 15  $\Theta \leftarrow \text{power\_control}(\mathcal{Q}_x);$ 16 Move DUs in  $\Theta$  from  $\mathcal{Q}_x$  to  $\mathcal{D}'_k$ ; 17 foreach  $u_i \in \mathcal{Q}_x \cap \mathcal{D}$  do 18 if  $\sigma_i^x < \sigma_i^{\min}$  then 19 Move  $u_i$  from  $\mathcal{Q}_x$  to  $\mathcal{D}'_k$ ; 20

## IV. THE PROPOSED DACS FRAMEWORK

DACS contains three core algorithms. In the *basic resource allocation (BRA)* algorithm, each SP allots its RBs to its UEs. If there are some UEs not served, the *inter-SP resource sharing (IRS)* algorithm enables resource lending among SPs. The *RB throughput improvement (RTI)* algorithm increases the number of data bits carried by RBs. Moreover, there are three auxiliary procedures. The *power control* procedure adjusts the power of each sender on the same RB. Both *SP prioritization* and *lender selection* procedures implement the RB lending mechanism.

## A. Core Algorithms

Given each SP  $s_k$ 's UEs, the BRA algorithm decides their initial RB and power allocation. Algo. 1 shows its pseudocode. Lines 1–4 check if UEs  $u_i$  and  $u_j$  are neighbors, that is,  $u_i$ 's sender,  $\varepsilon(i)$ , imposes non-neglected interference on  $u_j$ . To do so, we check if gain  $g_{\varepsilon(i),j}$  exceeds threshold  $g_{\text{th}}$ . In line 4, indicator  $\zeta_{i,j}$  reveals whether  $u_i$  and  $u_j$  are neighbors ( $\zeta_{i,j} = 1$ if so, or  $\zeta_{i,j} = 0$  otherwise; the default value of  $\zeta_{i,j}$  is zero).

The subsequent for-loop allots RBs to  $s_k$ 's CUs (i.e., lines 7–10) and DUs (i.e., lines 11–14) and finds their power (i.e., lines 15–20). BRA serves CUs with larger gains first, as their SINR demands can be met using lower power. So, line 6 sorts CUs in  $C_k$  by their gains from the BS decreasingly and stores the result in a set  $C'_k$ . Let  $Q_x$  be the set of UEs using RB  $r_x$ . For each CU  $u_i$  in  $C'_k$ , we choose an RB  $r_x$  from  $\mathcal{R}_k$  for it if two conditions both hold: 1)  $r_x$  is not assigned to any CU,

# Algorithm 2: Inter-SP Resource Sharing (IRS)

1 8	$C_{\mathbf{B}} \leftarrow \text{SP\_prioritization}();$		
2 fo	breach $s_k \in \mathcal{S}_{\mathbf{B}}$ do		
3	$\mathcal{S}_{\mathbf{L}} \leftarrow \texttt{lender\_selection}(s_k);$		
4	$ C'_k \leftarrow f_{\text{SRT}}^{\text{DEC}}(\mathcal{C}'_k, g_{\text{BS},i}), \ \mathcal{D}'_k \leftarrow f_{\text{SRT}}^{\text{DEC}}(\mathcal{D}'_k, g_{\text{BS},i}); $		
5	foreach $u_i \in \mathcal{C}'_k$ do		
6	foreach $s_m \in \mathcal{S}_{\mathbf{L}}$ do		
7	foreach $r_x \in \mathcal{R}_m$ do		
8	<b>if</b> $\mathcal{Q}_x \cap \mathcal{C} = \emptyset$ and $ \mathcal{Q}_x  < \delta$ and		
	$\zeta_{i,j} = 0, \ \forall u_j \in \mathcal{Q}_x$ then		
9	Move $u_i$ from $\mathcal{C}'_k$ to $\mathcal{Q}_x$ ;		
10	$\Theta \leftarrow \text{power\_control}(\mathcal{Q}_x);$		
11	$ \qquad \qquad$		
	then		
12	Move $u_i$ from $\mathcal{Q}_x$ to $\mathcal{C}'_k$ and		
	restore power of UEs in $Q_x$ ;		
13	else		
14	<b>go to</b> line 5;		
15	foreach $u_i \in \mathcal{D}'_{l_i}$ do		
16	foreach $s_m \in \mathcal{S}_{\mathbf{L}}$ do		
17	foreach $r_x \in \mathcal{R}_m$ do		
18	<b>if</b> $ \mathcal{Q}_x  < \delta$ and $\zeta_{i,j} = 0, \forall u_i \in \mathcal{Q}_x$		
	then		
19	Move $u_i$ from $\mathcal{D}'_k$ to $\mathcal{Q}_x$ ;		
20	$\Theta \leftarrow power\_control(\mathcal{Q}_x);$		
21	<b>if</b> $\Theta \neq \emptyset$ or $\sigma_i^x < \sigma_i^{\min}, \exists u_i \in \mathcal{Q}_x$		
	then		
22	Move $u_i$ from $\mathcal{Q}_x$ to $\mathcal{D}'_k$ and		
	restore power of UEs in $Q_x$ ;		
23	else		
24	<b>go to</b> line 15;		

and 2) there is no  $u_i$ 's neighbor that uses  $r_x$ . If  $r_x$  is found, we remove  $u_i$  from  $C'_k$  and add it to  $Q_x$ .

Then,  $s_k$  allots RBs to its DUs. BRA first serves DUs whose gains from the BS are larger, since they are more susceptible to the BS's interference. Line 6 sorts DUs in  $\mathcal{D}_k$  by gain  $g_{\text{BS},i}$ in a decreasing order, and the result is stored in a set  $\mathcal{D}'_k$ . In lines 11–14, we pick an RB  $r_x \in \mathcal{R}_k$  for each DU  $u_i$  in  $\mathcal{D}'_k$  if two conditions both hold: 1)  $r_x$  can accommodate UEs (i.e.,  $|\mathcal{Q}_x| < \delta$ ) and 2) there is no  $u_i$ 's neighbor also using  $r_x$ . After finding RB  $r_x$ , we remove  $u_i$  from  $\mathcal{D}'_k$  and add it to  $\mathcal{Q}_x$ .

The remaining code computes power for UEs that use each RB  $r_x \in \mathcal{R}_k$ . To do so, we use the power control procedure in line 16, which takes  $\mathcal{Q}_x$  as its input parameter. As discussed later in Section IV-B, this procedure returns a subset of DUs,  $\Theta$ , from  $\mathcal{Q}_x$  that cannot allocate power. Hence, line 17 removes these DUs from  $\mathcal{Q}_x$  and adds them back to  $\mathcal{D}'_k$ . The for-loop in lines 18–20 further excludes each DU from  $\mathcal{Q}_x$  whose SINR  $\sigma_i^x$  cannot meet its minimum requirement  $\sigma_i^{\min}$ .

## Algorithm 3: RB Throughput Improvement (RTI)

1 foreach $r_x \in \mathcal{R}$ do				
2	foreach $u_i \in \mathcal{Q}_x$ do			
3	$\left[ \begin{array}{c} { ilde g}_i^{ ext{sum}} \leftarrow \sum_{u_j \in \mathcal{Q}_x ackslash \{u_i\}} g_{arepsilon(i),j}; \end{array}  ight.$			
4	$\mathcal{Q}_x \leftarrow f_{\mathtt{SRT}}^{\mathtt{INC}}(\mathcal{Q}_x,  ilde{g}_i^{\mathtt{sum}});$			
5	foreach $u_i \in \mathcal{Q}_x$ do			
6	$\alpha \leftarrow 0,  \beta \leftarrow (p_i^{\max} - p_{\varepsilon(i),i}^x) / \alpha_{\max};$			
7	while $\alpha < \alpha_{\max}$ and $\beta > \beta_{\min}$ do			
8	$\Gamma_{\text{pre}} \leftarrow \sum_{u_j \in \mathcal{Q}_x} \gamma_j^x;$			
9	$p_{\varepsilon(i),i}^x \leftarrow p_{\varepsilon(i),i}^x + \beta,  \Gamma_{\texttt{alt}} \leftarrow \sum_{u_j \in \mathcal{Q}_x} \gamma_j^x;$			
10	<b>if</b> $\Gamma_{\texttt{alt}} \leq \Gamma_{\texttt{pre}} \text{ or } \sigma_j^x < \sigma_j^{\min}, \exists u_j \in \mathcal{Q}_x$			
	then			
11	$[p_{\varepsilon(i),i}^x \leftarrow p_{\varepsilon(i),i}^x - \beta; \text{ break};$			
12				

When an SP  $s_k$  has not enough RBs to serve its UEs (i.e.,  $C'_k \cup D'_k \neq \emptyset$ ), the IRS algorithm helps  $s_k$  borrow RBs from other SPs to serve these UEs. Algo. 2 shows IRS's pseudocode. Line 1 invokes the SP prioritization procedure to get a sorted set  $S_B$ , which determines the order of SPs to borrow RBs from others. Based on the order, the subsequent for-loop iteratively picks an SP  $s_k$  from  $S_B$ . Then, line 3 calls the lender selection procedure to get a sorted set  $S_L$  that decides the order of SPs to lend RBs to  $s_k$ , where  $S_L$  does not include  $s_k$ .

Like BRA, line 4 sorts CUs in  $\mathcal{C}'_k$  and DUs in  $\mathcal{D}'_k$  by their gains from the BS decreasingly. Then, we first handle CUs in  $\mathcal{C}'_k$  (in lines 5–14) and then DUs in  $\mathcal{D}'_k$  (in lines 15–24). For each CU  $u_i$  in  $\mathcal{C}'_k$ , we iteratively pick an SP  $s_m$  from  $\mathcal{S}_L$  and then find an RB  $r_x$  from  $\mathcal{R}_m$  (i.e., set of  $s_m$ 's RBs). Then,  $r_x$ is a candidate RB used to serve  $u_i$  if three conditions all hold: 1)  $r_x$  is not assigned to any CU (i.e.,  $Q_x \cap C = \emptyset$ ), 2)  $r_x$  still has room for serving  $u_i$  (i.e.,  $|Q_x| < \delta$ ), and 3) there are no  $u_i$ 's neighbors also using  $r_x$  (i.e.,  $\zeta_{i,j} = 0, \forall u_j \in Q_x$ ). If so, line 10 calls the power control procedure to recalculate power for UEs in  $\mathcal{Q}_x$  (including  $u_i$ ). However, if this procedure cannot allocate power to some UEs in  $Q_x$  (i.e.,  $\Theta \neq \emptyset$ ) or some UEs in  $Q_x$  cannot meet their SINR demands (i.e.,  $\sigma_j^x < \sigma_j^{\min}, \exists u_j \in$  $Q_x$ ), it is inappropriate to let  $u_i$  use  $r_x$ . In this case, we remove  $u_i$  from  $\mathcal{Q}_x$ , add  $u_i$  back to  $\mathcal{C}'_k$ , and restore the original power of all UEs in  $Q_x$ , as shown in line 12. Otherwise, we go back to line 5 to pick the next UE in  $C'_k$  to serve.

The way to handle DUs of  $\mathcal{D}'_k$  by lines 15–24 is similar to that to handle CUs, except that we check only two conditions,  $|\mathcal{Q}_x| < \delta$  and  $\zeta_{i,j} = 0, \ \forall u_j \in \mathcal{Q}_x$ , in line 18.

For each RB  $r_x \in \mathcal{R}$ , the RTI algorithm in Algo. 3 raises power for those UEs in  $\mathcal{Q}_x$  to improve  $r_x$ 's throughput. Since raising power would increase interference with other UEs, we prioritize raising power for UEs with less impact on other UEs. Lines 2–4 contain the code, where  $\tilde{g}_i^{\text{sum}}$  is the sum of gains on all UEs (except for  $u_i$ ) in  $\mathcal{Q}_x$  by  $u_i$ 's sender  $\varepsilon(i)$ . A lower  $\tilde{g}_i^{\text{sum}}$  value means that  $\varepsilon(i)$  has less impact. Hence, line 4 uses

**Procedure** power control  $(\mathcal{Q}_x)$ :  $\Theta \leftarrow \emptyset;$ 1 foreach  $u_i \in Q_x$  do 2 Estimate initial power  $p_i$  for  $u_i$ 's sender; 3 if  $u_i \in \mathcal{D}$  and  $p_i > p_i^{\max}$  then 4  $| \Theta \leftarrow \Theta \cup \{u_i\};$ 5  $\mathcal{Q}'_x \leftarrow \mathcal{Q}_x \setminus \Theta;$ 6 foreach  $u_i \in \mathcal{Q}'_r$  do 7  $\tilde{g}_i^{\text{sum}} \leftarrow \sum_{u_j \in \mathcal{Q}_x \setminus \{u_i\}} g_{\varepsilon(i),j};$ 8  $\mathcal{Q}'_x \leftarrow f_{\mathtt{SRT}}^{\mathtt{INC}}(\mathcal{Q}'_x, \tilde{g}_i^{\mathtt{sum}});$ 9 foreach  $u_i \in \mathcal{Q}'_x$  do 10  $\Lambda_i^{\text{LB}} \leftarrow \max\{p_i, p_i^{\min}\}, \Lambda_i^{\text{UB}} \leftarrow p_i^{\max};$ 11 12 
$$\begin{split} & \mathbf{P}_{\mathrm{L}} \leftarrow \boldsymbol{\Lambda}_{i}^{\mathrm{LB}} + (\boldsymbol{\Lambda}_{i}^{\mathrm{UB}} - \boldsymbol{\Lambda}_{i}^{\mathrm{LB}}) \times 1/4; \\ & \mathbf{P}_{\mathrm{M}} \leftarrow \boldsymbol{\Lambda}_{i}^{\mathrm{LB}} + (\boldsymbol{\Lambda}_{i}^{\mathrm{UB}} - \boldsymbol{\Lambda}_{i}^{\mathrm{LB}}) \times 1/2; \\ & \mathbf{P}_{\mathrm{H}} \leftarrow \boldsymbol{\Lambda}_{i}^{\mathrm{LB}} + (\boldsymbol{\Lambda}_{i}^{\mathrm{UB}} - \boldsymbol{\Lambda}_{i}^{\mathrm{LB}}) \times 3/4; \\ & \text{if } \boldsymbol{\sigma}_{i}^{x} \geq \boldsymbol{\sigma}_{i}^{\min} \text{ and decrement in } \boldsymbol{\sigma}_{i}^{x} < 50\% \end{split}$$
13 14 15 16 using  $P_L$  then  $p_{\varepsilon(i),i}^{\tilde{x}} \leftarrow \mathcal{P}_{\mathsf{L}}, \ \Lambda_{i}^{\mathsf{UB}} \leftarrow \Lambda_{i}^{\mathsf{UB}} \times 3/4;$ 17 else if increment in  $\sigma_i^x > 50\%$  using  $P_H$ 18 then  $p_{\varepsilon(i),i}^{x} \leftarrow \mathrm{P}_{\mathrm{H}}, \Lambda_{i}^{\mathrm{LB}} \leftarrow \Lambda_{i}^{\mathrm{LB}} \times 5/4;$ 19 else 20  $\begin{array}{l} p_{\varepsilon(i),i}^{x} \leftarrow \mathbf{P}_{\mathrm{M}};\\ \Lambda_{i}^{\mathrm{UB}} \leftarrow \Lambda_{i}^{\mathrm{UB}} \times 3/4, \ \Lambda_{i}^{\mathrm{LB}} \leftarrow \Lambda_{i}^{\mathrm{LB}} \times 5/4; \end{array}$ 21 22 while  $P_{H} - P_{L} \ge \rho$ ; 23 return  $\Theta$ ; 24

this value to sort UEs in  $Q_x$  increasingly.

For each UE  $u_i \in Q_x$ , its sender can increase power by at most  $p_i^{\max} - p_{\varepsilon(i),i}^x$ , which we divide into  $\alpha_{\max}$  equal parts (i.e.,  $\beta$ ) and  $\alpha_{\max} > 1$ . Line 7 presents two conditions for whether to run the while-loop: 1)  $\varepsilon(i)$ 's power is below its maximum value (i.e.,  $\alpha < \alpha_{\max}$ ) and 2) the adjustable power  $\beta$  is above threshold  $\beta_{\min}$  (otherwise, it means that  $\varepsilon(i)$ 's power is close to  $p_i^{\max}$ , and there is no point to increase  $\varepsilon(i)$ 's power). Then, we use  $\Gamma_{\text{pre}}$  and  $\Gamma_{\text{alt}}$  to represent  $r_x$ 's throughput before and after increasing  $\varepsilon(i)$ 's power by  $\beta$ , where  $\gamma_j^x$  is the number of data bits that a UE  $u_j$  can get from  $r_x$ . If raising  $\varepsilon(i)$ 's power has no benefit (i.e.,  $\Gamma_{\text{alt}} \leq \Gamma_{\text{pre}}$ ) or forces some UEs in  $Q_x$  to fail to meet their SINR demands (i.e.,  $\sigma_j^x < \sigma_j^{\min}, \exists u_j \in Q_x$ ), we should restore the previous power value for  $\varepsilon(i)$  and break the while-loop (as  $\varepsilon(i)$ 's power can no longer increase).

# B. Auxiliary Procedures

Given a set  $Q_x$  of UEs that share RB  $r_x$ , the power control procedure computes the amount of power for the sender  $\varepsilon(i)$ of each UE  $u_i$  in  $Q_x$  such that  $u_i$ 's SINR demand (i.e.,  $\sigma_i^{\min}$ ) is met while  $\varepsilon(i)$ 's interference can be reduced. This procedure will return a subset  $\Theta$  of DUs in  $Q_x$  whose senders cannot be allocated power. The for-loop in lines 2–5 estimates the initial power (denoted by  $p_i$ ) for  $u_i$ 's sender. There are two cases. If

Procedure SP\_prioritization ():

 1
 foreach 
$$s_k \in S$$
 do

 2
  $\Omega_k^{sum} \leftarrow 0;$ 

 3
 foreach  $r_x \in \mathcal{R}_k$  do

 4
  $\begin{bmatrix} foreach s_m \in S \setminus \{s_k\} do \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & &$ 

 $u_i$  is a CU, we consider that it is interfered by thermal noise  $\varphi$ . Hence,  $p_i$  can be derived as follows:  $(g_{BS,i} \times p_i)/\varphi \ge \sigma_i^{\min} \Rightarrow p_i = (\sigma_i^{\min} \times \varphi)/g_{BS,i}$ . When  $u_i$  is a DU, we take account of the BS's interference:

$$\frac{g_{\varepsilon(i),i} \times p_i}{g_{\text{BS},i} \times p_i^{\min} + \varphi} \ge \sigma_i^{\min} \Rightarrow p_i = \frac{\sigma_i^{\min}(g_{\text{BS},i} \times p_j^{\min} + \varphi)}{g_{\varepsilon(i),i}}.$$

However, if  $p_i > p_i^{\max}$ , it means that  $\varepsilon(i)$ 's power cannot be allocated, so we add  $u_i$  to  $\Theta$ . The code is given in lines 4–5.

Let  $Q'_x \subseteq Q_x$  denote the set of UEs whose senders' power are allocatable. Since raising power may increase interference with other UEs, we prioritize raising power for UEs with less effect. To do so, the for-loop in lines 7–8 computes the sum of gains  $\tilde{g}_i^{\text{sum}}$  on all UEs (excluding  $u_i$ ) in  $Q_x$  by  $u_i$ 's sender  $\varepsilon(i)$ , and line 9 sorts  $Q'_x$  increasingly by the  $\tilde{g}_i^{\text{sum}}$  value.

The next for-loop decides power for the sender of each UE  $u_i$  in  $\mathcal{Q}'_x$  based on a ternary search. Line 11 sets a power lower bound  $\Lambda_i^{\text{LB}}$  to the maximum of  $p_i$  (i.e., initial power) and  $p_i^{\min}$ (i.e., minimal power) and a power upper bound  $\Lambda_i^{\text{UB}}$  to  $p_i^{\text{max}}$ . In addition, we set three power values  $\mathrm{P}_L,\,\mathrm{P}_M,$  and  $\mathrm{P}_H,$  located at 1/4, 1/2, and 3/4 of the range from  $\Lambda_i^{\text{LB}}$  to  $\Lambda_i^{\text{HB}}$ , in lines 13-15. Three cases are used to adjust power. In case 1 (i.e., lines 16–17),  $u_i$  meets its SINR demand using a lower power, and the decrement in  $\sigma_i^x$  is less than a half. Hence, it is safe to set  $p_{\varepsilon(i),i}^x = P_L$ , and we lower the power upper bound  $\Lambda_i^{UB}$ . In case 2 (i.e., lines 18–19), using a high power improves  $u_i$ 's SINR, so we set  $p_{\varepsilon(i),i}^{x}$  to  $P_{H}$  and raise the power lower bound  $\Lambda_i^{\text{LB}}$ . Otherwise (i.e., case 3 in lines 20–22), we set  $p_{\varepsilon(i),i}^x$  to  $P_{M}$  (i.e., the medium power) and shrink the range between  $\Lambda_{i}^{LB}$ and  $\Lambda_i^{\text{UB}}$ . The above iterations are repeated until the difference between  $P_{\rm H}$  and  $P_{\rm L}$  is below a threshold  $\rho$ .

The SP prioritization procedure finds the order of SPs that can borrow RBs from others. Given an RB  $r_x$  in  $\mathcal{R}_k$ , we define the resource amount of  $r_x$  that  $s_k$  lends to another SP  $s_m$  by  $\Omega_{k,m}^x = (|\mathcal{Q}_x \cap (\mathcal{C}_m \cup \mathcal{D}_m)|)/|\mathcal{Q}_x|$ . Specifically, the numerator is the number of  $s_m$ 's UEs using  $r_x$  and the denominator gives Procedure lender\_selection  $(s_k)$ : foreach  $s_m \in S \setminus \{s_k\}$  do  $\Psi_m \leftarrow 0;$ foreach  $r_x \in \mathcal{R}_k$  do  $\Psi_m \leftarrow \Psi_m + \Omega_{k,m}^x;$ freturn  $f_{\text{SRT}}^{\text{DEC}}(S \setminus \{s_k\}, \Psi_m);$ 

the number of total UEs that share  $r_x$ . Then, this procedure has two modes:

- 1. Lines 1–5 and 11–12: When  $s_k$  has lent more resources to others (i.e.,  $\Omega_k^{\text{sum}}$  is larger), it has a higher priority to borrow RBs. This mode is to encourage SPs to lend RBs.
- 2. Lines 6–10 and 13: If  $s_k$  has borrowed fewer resources from other SPs (i.e.,  $\Psi_m^{\text{sum}}$  is smaller), it possesses a higher priority to borrow RBs. This mode is for fairness.

We use mode 1 by default and periodically switch modes.

Given SP  $s_k$ , the lender selection procedure finds the order of other SPs that lend RBs to  $s_k$ . Generally, if an SP  $s_m$  has borrowed more resources from  $s_k$ ,  $s_m$  shall lend back its RBs to  $s_k$  first. In this way, we can better maintain fairness among SPs. Except  $s_k$ , for every SP  $s_m$  in S, the for-loop in lines 1–4 computes the total resource amount (denoted by  $\Psi_m$ ) that  $s_m$ has borrowed from  $s_k$ . Then, line 5 sorts all SPs in  $S \setminus \{s_k\}$ decreasingly by their  $\Psi_m$  values and returns the sorted result.

## V. EXPERIMENTAL EVALUATION

Our simulation is built on MATLAB, whose parameters are given in Table I. One BS offers 45 RBs/ms, where SPs  $s_1$ ,  $s_2$ , and  $s_3$  possess 10, 15, and 20 RBs. Each SP needs to serve 10 CUs and 5 to 30 D2D pairs. The minimum power for the BS and a D2D sender is set to -40 dBm. For path loss,  $L(\cdot, \cdot)$  is the distance between a sender and a receiver, as measured in km. Shadowing fading is modeled using a log-normal distribution [21], with a mean and standard deviation of 0 dB and 8 dB.

We pick two methods for comparison. The *wireless resource virtualization with D2D communication (WRVD)* method [15] allocates RBs to CUs for different SPs. Each RB is reused by a D2D pair with the minimum interference. The *joint resource allocation and power control with RAN sharing (JAPS)* method [19] adjusts senders' power and lets SPs borrow RBs from others to improve throughput. We also consider a reduced version of our DACS framework (called *R-DACS*), a combination of only BRA and RTI algorithms. By comparing DACS with R-DACS, we can evaluate the impact extent of the IRS algorithm. In JAPS, R-DACS, and DACS, we set  $\delta = 30$ .

## A. Comparison on Performance

Fig. 1(a) compares the service rate. In general, the service rate reduces when the number of D2D pairs grows, since more DUs compete for the fixed resource. WRVD makes each RB be reused by at most a D2D pair. Once there are more D2D pairs, many DUs cannot be allocated RBs. Therefore, WRVD has the lowest service rate. R-DACS allows an RB to be shared by at

TABLE I SIMULATION PARAMETERS.

BS parameters:				
cell range	500 m (channel bandwidth: 180 kHz)			
maximum power	$46 \mathrm{dBm} \ (\approx 40 \mathrm{W})$			
RBs per ms	45 (SP s <sub>1</sub> , s <sub>2</sub> , s <sub>3</sub> own 10, 15, 20 RBs)			
UE parameters:				
number of UEs	CUs: 30, D2D pairs: 15, 30, 45, 60, 75, 90			
maximum power	$23 \mathrm{dBm} \ (\approx 0.2 \mathrm{W})$			
throughput demand	1 Mbps (i.e., $\sigma_i^{\min} \approx 17  \text{dB}$ )			
Channel parameters:				
path loss	BS to UE: $128.1 + 37.6 \log_{10} L(BS, u_i)$			
-	UE to UE: $148 + 40 \log_{10} L(\varepsilon(i), u_i)$			
shadowing	log-normal distribution			
thermal noise	-174 dBm/Hz			



Fig. 1. Comparison of performance in WRVD, JAPS, R-DACS, and DACS.

most  $\delta$  UEs ( $\delta = 30$ ), thereby improving RB utilization and the service rate. When an SP has insufficient RBs to serve its UEs, JAPS and DACS make the SP borrow RBs from others. They achieve nearly 0.98 of a service rate. This result demonstrates that both JAPS and DACS can exploit RAN sharing to increase the service rate significantly.

In Fig. 1(b), we evaluate energy efficiency. As WRVD has the lowest service rate and adopts fixed power for the BS and D2D senders, its energy efficiency is greatly lower than other methods. Though JAPS has a similar service rate with DACS, its energy efficiency is significantly lower than that of DACS. This result reflects that DACS performs much better than JAPS in power control.

# B. Comparison on Fairness

Both JAPS and DACS support RB lending mechanisms to let SPs borrow resources from each other to improve the service rate. In this section, we judge whether their mechanisms can maintain fairness among SPs.

Fig. 2(a) and (b) show each SP's throughput when there are 30 and 90 D2D pairs. Among 45 RBs offered by the BS (per ms), SPs  $s_1$ ,  $s_2$ , and  $s_3$  own 10, 15, and 20 RBs, which occupy 22.22%, 33.33%, and 44.45% of the BS's resources. Evidently, the gap between  $s_1$ 's throughput and  $s_3$ 's throughput is much larger in JAPS than that in DACS. Fig. 2(c) exhibits the ratio of each SP's throughput. Compared to JAPS, this ratio in DACS can be closer to the ratio of SPs' owned resources.

Fig. 3 gives the resource amount that each SP borrows and lends. JAPS will assign a high priority to an SP that has lent more RBs to others. Since SP  $s_3$  owns the most RBs, it has a good chance of loaning out more RBs in the early stage. This



(c) ratio of each SP's throughput  $(s_1 : s_2 : s_3)$ 

Fig. 2. Comparison of each SP's throughput.



Fig. 3. Comparison of resource amount that each SP borrows and lends.

makes  $s_3$  have a high priority for a long time. Eventually,  $s_3$  will borrow much more RBs than it has lent. On the contrary, SP  $s_1$  is forced to lend much more RBs than it borrows. These phenomena verify the unfairness of JAPS in RB borrowing.

In DACS, the SP prioritization procedure has two modes: 1) assigning a high priority to an SP that lent more resources, and 2) assigning a high priority to an SP that borrowed fewer resources. By switching between these two modes periodically, DACS can efficiently deal with the above problem. Moreover, the lender selection procedure asks an SP that has borrowed more resources from SP  $s_k$  to first repay RBs to  $s_k$ . As can be seen from Fig. 3, the resource amount that each SP borrows is similar to that it lends, which verifies that DACS can maintain fairness among SPs.

# VI. CONCLUSION

This paper proposes the DACS framework to handle D2D resource allocation and power control with RAN sharing. Each SP allots its RBs to UEs and borrows RBs from others when it has not enough RBs. Then, the power of senders is raised for throughput improvement. In DACS, the power control procedure uses a ternary search to find suitable power. Moreover, the RB lending mechanism considers balancing between efficiency and fairness. Through simulations in MATLAB, we show that DACS not only attains a high service rate but also increases energy efficiency, as compared to WRVD, R-DACS, and JAPS. Moreover, DACS can keep fairness among SPs.

## ACKNOWLEDGMENT

This work was supported by National Science and Technology Council, Taiwan under Grant 113-2221-E-110-056-MY3.

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