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A spatial slotted-Aloha protocol in wireless networks for group communications

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Abstract

In a large-scale group communication networks, we propose a slotted-Aloha-based access control (SA-AC) scheme with the optimal transmission probability (TP) using the stochastic geometry. Since the performance of SA-AC scheme depends on the TP of the members in a group, the analytical model presents a joint view for the downlink (DL) and the uplink (UL) with TP of the members in a group. We present simple and closed form expressions for the DL and UL joint probability by using Poisson point process (PPP). Furthermore, we analyze the dynamic TP and the optimal TP to maximize the DL/UL joint probability, for which the inclusion of a member and the transmission of the member is determined by the DL and UL thresholds. Since the requirement of the DL/UL joint probability varies with the type of services, the optimal TP has to be determined for an efficient group communication service. The performance of the SA-AC scheme with the optimal TP is demonstrated, and it is shown to be superior over the other schemes.

Keywords: Group communication, Spatial Aloha, Stochastic geometry

1 Introduction

In a large-scale group communication network, some nodes may form a cluster and clusters communicate with one another. For example, in a tactical group communication network, a commander monitors the conditions of the soldiers. A soldier equips with a monitoring system [1], and the monitored status is transmitted to the base camp through an unmanned aerial vehicle (UAV) [2]. If all the soldiers transmit their information at the same time to the UAV, the network congestion will occur increasingly as the number of soldiers increases. Thus, a hierarchical system in which relays or vehicles collect the information of the soldiers and they transmit the collected information to the UAV could be desirable. In the system, the soldiers are grouped to communicate with a relay or a vehicle, and they act as the group members and the relay or the vehicle becomes the group leader. The group leader has to cover the group members as many as possible within its communication range. The members have to choose the best group leader among the candidate group leaders. The group leader will control the transmissions of the group members within its coverage.

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With a massive number of nodes, the group leaders and the members form a large-scale network and they can be modeled by the Poisson point process (PPP). PPP is a spatial point process, which is widely used to model a wireless network. However, a large-scale network with group communications in which group leaders and members send and receive information has not been studied. Since the members within the coverage of a leader are able to transmit for communications, the locations of the members are no longer independent to the location of the leader. Moreover, the interference model should be different from that in [3] because the interference from the covered area of a leader is important compared to that from the noncovered area of a leader. In addition, the covered members of a group may decide whether to transmit or not by a transmission control (TC) value, which is globally determined by the leaders in a centralized manner. Thus, a new network model to capture the behavior of hierarchical group communications needs to be setup.

Since wireless resources in a group communication network are shared in a distributed manner, the performance depends on the medium access control (MAC) scheme. In this sense, transmission probability (TP), which is decided by the TC value, should be carefully chosen. Since the TP of nodes in an Aloha-based



scheme may be adaptively determined by channel variation, local topology, and target utility, a new MAC protocol can be designed by using the optimal TP which is derived from the PPP models in group communication networks.

In this paper, we first derive the dynamic TP which is determined by the statistical distributions of group leaders and the DL coverage probability. We focus on the special case to derive the closed-form expressions. The coverage probability and the average numbers of members per leader in the DL are derived for the dynamic TP. The dynamic TP has been known to be optimal in terms of the MAC layer. However, the performance using the dynamic TP has not been evaluated by considering geometrical distributions of network elements.

Next, we propose the optimal TP which maximizes the downlink (DL) and uplink (UL) joint probability. The DL/UL joint probability is that the transmission of the leader and the transmission of a member are performed within the DL coverage threshold and the UL transmission threshold. The probability is determined by the probability density function (pdf) of the distance between the leader and a member which is within the coverage of the leader. The pdf is derived and is utilized to model the UL probability and the DL/UL joint probability. Since the DL/UL joint probability is maximized for a certain distance between the leader and a member, the closedform expressions of the optimal TP for the distance can be derived. In a practical system, the distance is usually hard to be determined. In our proposed scheme, however, a leader can determine the optimal TP by using the average number of members per leader.

Finally, we present the performance of the DL/UL joint probability and the average achievable rate for a target distance in our model. In addition, the average achievable rates for the different TPs are provided. From the analytical model, the effect of TP in a group communication network is provided. Furthermore, a new policy to determine the optimal TP is proposed by considering the geometrical effects of network elements. For the group communication network, the performance at a target distance is maximized by the proposed slotted-Aloha-based access control (SA-AC) scheme.

The contributions of our work are summarized in the following points:

- We propose a new SA-AC scheme with the optimal TP to maximize the DL/UL joint probability for group communication network.
- We define pdf of the distance between a leader and its covered member for a given DL threshold, and the closed form expression of the optimal TP to maximize the DL/UL joint probability for a target member.

 We evaluate the performance of the proposed SA-AC scheme with the traditional schemes.

The rest of this paper is organized as follows. Section 2 reviews the recent PPP studies of DL, UL, and the MAC perspectives. In Section 3, we explain the proposed SA-AC scheme and the other SA-AC schemes in group communication networks. Section 4 analyzes the dynamic TP and the optimal TP by using the PPP model and presents an access scheme with the optimal TP. Section 5 presents numerical results to demonstrate the access control schemes with different TPs. Finally, we conclude in Section 6.

2 Related works

Modeling of the DL of a large-scale network using stochastic geometry has been studied in [4, 5]. The performance of the DL system depends on the locations of base stations (BSs), and it has been derived by modeling BSs as a PPP [6-9]. The general expressions for the coverage probability and achievable rate at a typical user are presented in [4]. In some special cases, the closedform expressions, which give an intuitive view for the DL system, are provided. In [5], the association probability for a certain type of BSs which are modeled as multiple PPPs with different spatial densities and system parameters is studied. The analytical model in [5] presents the outage and spectral efficiency performance for the DL system deploying different types of BSs. Most PPP models for the DL system are analyzed from the view point of a typical user with a DL communication threshold, which shows that the tractable models are fairly accurate compared to actual systems. The performance of the resource management schemes in the DL (e.g., coverage expansion [5], fractional frequency reuse (FFR) [10], and resource partitioning [11]) is analyzed by using the PPP.

The analytical model using PPP for the UL of a largescale network has also been developed. The models in [12–14] are developed by dividing Voronoi cells from the perspective of users which are modeled as a PPP. Each user has its own BSs and UL resources for transmission. The transmission power is controlled by themselves according to its locations. For a typical transmission, the other transmissions are assumed as interference. In another approach for the UL model in [3], cells are divided with respect to BSs. The set of active users which satisfy the cutoff threshold for the UL is able to communicate with their BSs. The cutoff threshold for the UL is the average received power required at the serving BS. For a typical active user operating on an allocated channel in its serving BS, the other active users on the channel in the other BSs are assumed to be as interferences. The model provides tractability by assuming that the interfering users constitute a PPP. Since the correlations among users occur by the locations of the associated BSs and the tagged channel, the assumption is required to provide the insights for the cellular UL performance. The performance of the UL with a resource management scheme also has been analyzed with the PPP as in the DL [13].

For the clustered network scenario, [15] and [16] is researched. In these works, clusters are formed based on parent PPP points. To model the clustered networks Mattěrn cluster process (MCP) and Thomas cluster process (TCP) is used. The interference from the inter and the intra clusters are separately analyzed and approximated using the clustered process properties. The groups or clusters are already formed as a point process without the DL/UL joint model. And the process of forming clusters is not considered when leaders and members are independently deployed.

In addition, the stochastic geometry modeling has been developed by focusing on MAC schemes, especially Aloha-based schemes [17-19]. Since a node in an ad hoc network may operate as a transmitter or a receiver, an analytical model using PPP can be classified by the model of a receiver [6]. If each transmitter node has its corresponding receiver, the model is called as "Poisson bipolar model" [20–23]. In the Poisson bipolar model, the transmitters are modeled as a PPP and are divided into two non-overlapping subsets, whether a transmitter does its role at a specific time or not by using a TP. If the transmitter and receiver are independently modeled as PPPs, the model is called as "independent receiver model." In the model, a transmitter is controlled by TP and the transmission of the transmitter can be received by all subsets of the receivers [6]. When the transmitter is modeled as a PPP and the receiver is modeled as the subset of transmitters by using a specific condition, the model is called as "mobile ad hoc network model." If the set of active transmitters is classified by TP, the set of inactive transmitters at a slot is the set of receivers [24]. Note that a source node has to send data to its far destination node in a mobile ad hoc network. Then, the intermediate nodes are required to relay data to its destination node. If some of the inactive transmitters are in the direction from a source node to a destination node, they can be the receivers of the transmitter, i.e., the source node. In [22], the optimal TPs are derived for the maximum throughput medium access, the max-min fairness medium access, and the proportional fairness medium access. In [21], the performance of the Aloha protocol which is optimized for the proportional fair medium access is analyzed. The authors in [24] propose a spatial reuse Aloha protocol for a multihop network with the optimal TP which maximizes the density of progress, the mean total distance traversed in one hop by all transmissions initialized in some unit area. However, the optimal TP for the group communications with the DL coverage and the UL transmission has not been studied.

3 Slotted-Aloha-based access control schemes in group communication networks

In this section, we introduce the proposed SA-AC scheme which utilizes the optimal TP maximizing the DL/UL joint probability for the target distance. The target distance can be varied by the policy of the proposed SA-AC scheme. If the policy of the proposed SA-AC scheme is to maximize the coverage, the optimal TP is determined for maximizing the performance of the outermost members. In the example, the interference for the outermost members should be decreased to satisfy the UL threshold. The distance between the leader and the outermost member is affected by the DL threshold and the interference from the other leaders. Furthermore, since the number of members in the coverage of the leader increases by extending the coverage, the interference should be controlled by TP. So the proposed SA-AC scheme computes the optimal TP which is determined by considering the DL threshold, the UL threshold, the density of the leaders, the density of the members, and the target distance.

We present a network model for group communications and SA-AC schemes with various TP. The notations are summarized in Table 1. Both the group leaders and the group members are positioned according to PPPs. The leaders are distributed according to a PPP Φ_l with λ_l , the intensity of the leaders per unit area. The members are arranged according to a PPP Φ_m with λ_m , the intensity of the members per unit area. In the network model, let us define a link from a leader to a member as DL and a link from a member to a leader as UL. In the DL, we assume that a typical member is located at the origin. The typical member in the DL is denoted as m, and it is associated with the nearest leader among the leaders. The leader at the origin in the UL is denoted as *l*. Since the distributions are invariant, the locations of the leaders for the typical member in the UL are the same as those in the DL. Let l_i denote the leader *i* and m_i be the member *i* where $l_i \in \Phi_l \setminus l$ and $m_i \in \Phi_m \setminus m$.

We assume an interference-limited network in which the signal-to-interference ratios (SIRs) both in the DL and the UL are considered. The transmission powers of leaders and members are P_l and P_m , respectively. The signals in both links experience the path loss. The path loss exponent is denoted as α . The Rayleigh fading with the unit average power is assumed between any points. Let $h_{x,y}$ be the random variable to represent the channel effect from x to y. Thus, the SIR at the receiver with the distance r in the DL from l to m (SIR at a typical member in the downlink) is given by

$$SIR_d(r) = \frac{P_l h_{l,m} r^{-\alpha}}{I_l},\tag{1}$$

Table 1 Summary of notations

Notation	Description
1	Leader at the origin in the uplink
I_i	Leader i
m	Typical member in the downlink
m_i	Member i
m_c	Set of covered members
Φ_I	PPP of leaders
Φ_m	PPP of members
Φ_{m_c}	Set of locations of covered members
$h_{x,y}$	Channel effect from x to y
λ_I	Intensity of leaders
λ_m	Intensity of members
P_I	Transmission power of leaders
P_m	Transmission power of members
α	Pathloss exponent
$p_{d,u}$	Downlink, uplink joint probability
p_u	Uplink coverage probability
Pd	Downlink coverage probability
T_d	Downlink threshold
T_u	Uplink threshold
p _a	Access control probability
r	Distance between m and its nearest leader
$r_{d,i}$	Distance between a leader i and a typical member
$r_{u,i}$	Distance between a member i and a typical leader
$r_{du,i}$	Distance between a covered member i and a typical leader
r _{tar}	Target distance
r̄ _{max}	Average maximum distance between l and its m_c
11	Cumulative interference from the leaders
I _m	Cumulative interference from the members
I_{m_c}	Cumulative interference from the covered members
τ	Transmission probability (TP)
$ au^{dyn}$	TP of the dynamic framed SA-AC
$ au^{\mathit{fix}}$	TP of the fixed framed SA-AC
K	Frame size
K ^{dyn}	Frame size of the dynamic framed SA-AC
K ^{fix}	Frame size of the fixed framed SA-AC
SIR_d	SIR at a typical member in the downlink
SIR _{II}	SIR at a typical leader in the uplink
$SIR_{d,u}$	SIR at a typical leader in the joint DL/UL transmission
\bar{N}_m	Average number of the covered members per leader
	Average number of the covered members per leader Average achievable rate of covered members
Ear	Average achievable rate of covered members

where I_l is the cumulative interference from the leaders except the serving leader,

$$I_{l} = \sum_{l_{i} \in \Phi_{l} \setminus l} P_{l} h_{l_{i},m} r_{d,i}^{-\alpha}, \tag{2}$$

where $r_{d,i}$ is the distance between a leader i and a typical member. Since r is invariant in the DL and the UL, the SIR at a typical leader in the UL is

$$SIR_{u}(r) = \frac{P_{m}h_{m,l}r^{-\alpha}}{I_{m}},$$
(3)

where I_m is the cumulative interference from the members except the member m. In the UL, I_m is given by

$$I_m = \sum_{m_i \in \Phi_m \setminus m} P_m h_{m_i, l} r_{u, i}^{-\alpha}, \tag{4}$$

where $r_{u,i}$ is the distance between a member i and a typical leader.

The SA-AC scheme is assumed to work in a group communication network where the time slots are synchronized. One viable solution may be equipping global positioning system (GPS) modules on the leaders and the members since the signals from the GPS satellites provide the rather accurate timing and location information. There may be also several methods which can provide the synchronization to the members without GPS modules. Once a network is grouped into several clusters, all nodes can be bounded by the parentchildren relationship. All members can be synchronized to the leader by periodically exchanging synchronization and acknowledgment packets with their corresponding leaders using pair-wise synchronization method [25, 26]. The location information may be piggybacked at the synchronization and acknowledgment packet for simultaneous time synchronization and location information distribution. Depending on the velocity of the nodes, the period between the synchronization and the location information distribution can be adjusted accordingly.

The time is divided into multiple slots, and they are partitioned into control slots and data slots. The leaders and the members transmit their data with the SA-AC scheme. A leader broadcasts a control frame in a synchronized control slot. The control frame of a leader includes the information for the group of the leader, the address of the leader, and the TC value. We assume that the TC value is shared by using a specific channel and they are identical among leaders. The control frame is decodable for a member when the SIR value is greater than the DL threshold, T_d . The member in the coverage of the leader is denoted as the covered member. Only the covered member joins the group of the nearest leader when the SIR value for the received control frame from the nearest leader is greater than T_d . Once the covered member decodes the control frame successfully, it gets the information to transmit its data frame to the leader in the UL. In the UL, the covered member transmits data frame in a data slot by using a TP. The TP is determined by the TC value and the type of the TC value depends on the type of the AC scheme. If the SA-AC scheme uses the access control probability, p_a , as the TC value, the TP is given by

$$\tau = p_a. (5)$$

In the SA-AC scheme using p_a , the covered members access each slot with p_a . If the SA-AC scheme uses the frame size as the TC value, the TP is given by

$$\tau = \frac{1}{K},\tag{6}$$

where *K* is the frame size which represents the number of data slots. In the SA-AC scheme using K, a covered member selects and accesses a slot among the K slots. In general, 1/K is modeled as the access probability of the contenders in a slot. The framed SA-AC scheme consists of the fixed framed SA-AC scheme and the dynamic framed SA-AC scheme. The fixed framed SA-AC scheme utilizes the fixed TP which is determined by the fixed frame size [27]. Let τ^{fix} and K^{fix} denote the TP and the frame size of the fixed framed SA-AC, respectively. The fixed framed SA-AC scheme utilizes $\tau^{fix} = 1/K^{fix}$. The dynamic framed SA-AC scheme utilizes the dynamic TP determined by the frame size which is dynamically changed according to the average number of the covered members per leader. The frame size in the dynamic framed SA-AC scheme is known to be optimal when the frame size is equal to the number of contenders [18, 28]. Let τ^{dyn} and K^{dyn} denote the TP and the frame size of the dynamic framed SA-AC scheme, respectively. Then the dynamic framed SA-AC scheme utilizes $\tau^{dyn} = 1/K^{dyn}$.

However, if the channel effect and pathloss are not considered, such dynamic TP may not be optimal. So we need to find a new optimal TP maximizing the performance of group communications network. The optimal TP is affected by λ_l , λ_m , T_d , and T_u , where T_u is the UL threshold. In addition, it has to be determined by considering the performance of the covered members. The performance varies by the distance between a leader and the members. In Fig. 1, an example of group communications network is shown. Fig. 1a presents the DL transmissions and Fig. 1b shows the UL transmissions. In DL, the leaders broadcast the control frames to the members. λ_l determines the distance between a member and its nearest leader, and T_d decides whether the member becomes the covered member. For example, in each group, three members near the leader become the covered members. These members receive the leader's control frames without channel error. Let the member with target distance r_{tar} denote the target member. The target member in group 2 receives interference signal from the leaders in groups 1 and 3, but this interference signal is tolerable for reception.

In UL, the transmission of a member occurs when the member is in the coverage of the leader and determines whether the covered member participates in the UL transmission by τ and T_u . The target member in group 2 transmits data to the leader. The leader in group 2 receives interference signal from the member in group 2 as well as the members in groups 1 and 3. To decode the UL transmission of the target member, the interference signals have to be controlled by the optimal TP τ^* which is determined by λ_l , λ_m , T_d , T_u , and r_{tar} .

4 Optimal transmission probability for group communications

In this section, we derive the dynamic TP and the optimal TP. To derive the dynamic TP, we need the DL coverage probability and the average number of the covered members per leader. For the optimal TP, we need to derive the UL coverage probability and the DL/UL joint probability for a target distance. The target distance is the distance between a leader and a target member for which the DL/UL joint probability is maximized. Finally, we develop an analytical model for the performance of the SA-AC schemes using the dynamic TP and the optimal TP. From the model, we derive an optimal TP of the SA-AC scheme.

In our network model, both the leaders and the members are assumed to be arranged according to PPPs. Therefore, the pdf of the distance *R* between a member and its nearest leader [4] can be expressed as

$$f_R(r) = 2\pi r \lambda_l \exp\left(-\pi r^2 \lambda_l\right),\tag{7}$$

where r is the distance between a member and its nearest leader. The members are divided into the covered members and the non-covered members by T_d . The following lemma provides the probability that a member is in the coverage of the nearest leader.

Lemma 1 For the DL threshold T_d , the DL coverage probability that a member with the distance r from its nearest leader is in the coverage of the leader is

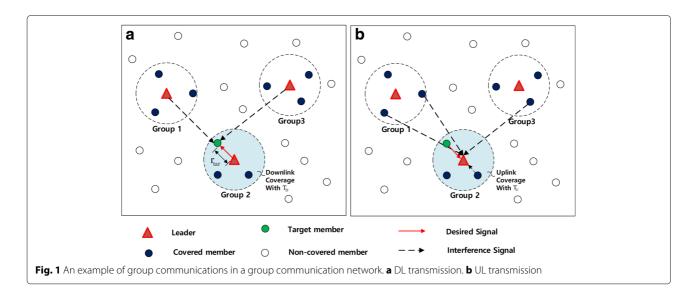
$$p_d(r, T_d) = e^{-\pi r^2 \lambda_l \zeta_l(T_d)}, \tag{8}$$

where $\zeta_l(T_d)=T_d^{2/\alpha}\int_{T_d^{-2/\alpha}}^\infty\frac{1}{1+(u_l)^{\alpha/2}}\mathrm{d}u_l$. Then, the DL coverage probability that a member is in the coverage of the leader is

$$p_d(T_d) = \frac{1}{1 + \zeta_l(T_d)}. (9)$$

Proof See Appendix A.1.
$$\Box$$

Lemma 1 shows that $p_d(r, T_d)$ decreases by increasing r, T_d , and λ_l . Since the received signal strength from the nearest leader decreases as r increases, $p_d(r, T_d)$ decreases. For a certain r, the SIR value may not be higher than T_d for large T_d . Since the increase of λ_l causes the increase of interference for a certain r, $p_d(r, T_d)$ decreases.



Lemma 1 implies that the number of the covered members decreases by increasing T_d for both $p_d(r,T_d)$ and $p_d(T_d)$. The expressions in Lemma 1 yield a closed-form when $\alpha=4$, since $\zeta_l(T_d)=\sqrt{T_d}\left(\frac{\pi}{2}-\arctan\left(\frac{1}{\sqrt{T_d}}\right)\right)$. From Lemma 1, we derive the dynamic TP, which

From Lemma 1, we derive the dynamic TP, which determines the transmission probability of the covered members.

Theorem 1 For the DL threshold T_d , the dynamic TP that is dynamically changed by the number of the averaged covered members is

$$\tau^{dyn} = \min\left(\frac{1}{\bar{N}_m}, 1\right) = \min\left(\frac{\lambda_l}{\lambda_m p_d(T_d)}, 1\right). \quad (10)$$

Proof If we assume that S is the area of an entire network, \bar{N}_m is obtained by dividing the average number of the covered members in S by the average number of the leaders in S. Since the intensity of the covered members is $\lambda_m p_d(T_d)$, the average number of the covered members per leader $\bar{N}_m = \frac{\lambda_m p_d(T_d)S}{\lambda_l S} = \frac{\lambda_m p_d(T_d)}{\lambda_l}$ [5].

Theorem 1 shows that τ^{dyn} is adaptive to the number of the covered members per leader. We expect that τ^{dyn} decreases as \bar{N}_m increases. It implies that the interference to the transmission of a typical covered member decreases as \bar{N}_m increases. Since \bar{N}_m is affected by $p_d(T_d)$, it increases by decreasing T_d . Thus, Theorem 1 indicates that τ^{dyn} decreases and the interference to the transmission of a typical covered member decreases as T_d decreases in the dynamic framed SA-AC scheme.

Since the locations of the covered members are jointly changed by T_d and the locations of the leaders, they are not PPP. It is challenging to model it accurately, but PPP approximation could be utilized [3]. In our model, we

approximate them as a PPP. Let Φ_{m_c} and Φ_{m_o} denote the locations of the covered members and the locations of the non-covered members. We calculate the pdf of the distance between a leader and a covered member from the following theorem.

Theorem 2 The pdf of the distance between a leader and a covered member m_c for a given T_d is

$$f_{R_{m_c}}(r) = \frac{2\pi r \lambda_l e^{-\pi r^2 \lambda_l (1 + \zeta_l(T_d))}}{p_d(T_d)}.$$
 (11)

Proof Since the coverage of a leader determines the distance between a leader and a covered member, the pdf is affected by the T_d . In our network model, the covered members are only able to transmit to their leader. Since the distribution of the distances between the leader and the members in DL is not changed in UL, the distribution of the distances between the leader and the covered members is induced by the pdf of R. In the UL transmission by using a TP, the interference from the covered members is obtained by normalizing the DL coverage probability at r by $p_d(T_d)$. Thus, $f_{R_{mc}}(r)$ is given by

$$f_{R_{m_c}}(r) = \frac{p_d(r, T_d)f_R(r)}{p_d(T_d)}.$$
 (12)

Plugging $p_d(r, T_d)$ in (8) into (12), $f_{R_{m_c}}(r)$ is derived. \square

Since $f_R(r)$ is conditioned by the DL coverage probability, $f_{R_{m_c}}(r)$ is affected by T_d . We expect that the increase of T_d makes the pdf of R_{m_c} within the DL coverage to increase. When $\alpha=4$, both $p_d(r,T_d)$ and $p_d(T_d)$ have closed-form expressions, and $f_{R_{m_c}}(r)$ has also a closed-form expression. From Theorem 2, we derive the UL probability that the SIR value for the transmission of a

covered member is larger than the UL threshold T_u . The SIR value in UL after DL is given by

$$SIR_{d,u}(r) = \frac{P_m h_{m_c,l} r^{-\alpha}}{I_{m_c}},$$
(13)

where I_{m_c} and $SIR_{d,u}(r)$ are the cumulative interference from the covered members except for the m_c at l and the SIR value for the transmission of the m_c at l with the distance r (SIR at a typical leader in the joint DL/UL transmission), respectively. In UL, I_{m_c} is given by

$$I_{m_c} = \sum_{m_{c,i} \in \Phi_{m_c} \setminus m_c} P_m h_{m_{c,i},l} \left(r_{du,i}^{-\alpha} \right), \tag{14}$$

where $r_{du,i}$ is the distance between a covered member i and a typical leader.

Lemma 2 For τ , the UL coverage probability that the SIR value for the transmission of the covered member with the distance r from its leader exceeds the UL threshold T_u is

$$p_u(r, \tau, T_u) = \exp\left(-\pi r^2 \lambda_m p_d(T_d) \zeta_m(T_u) \tau\right), \qquad (15)$$

where $\zeta_m(T_u) = T_u^{2/\alpha} \frac{2\pi/\alpha}{\sin(2\pi/\alpha)}$.

Lemma 2 shows that $p_u(r, \tau, T_u)$ decreases by increasing r, T_u , and λ_m . Since the received signal strength from the covered member decreases as r increases, $p_u(r, \tau, T_u)$ decreases. For a certain r, the SIR value for the signal may be hard to exceed T_u for large T_u . Since the increase of λ_m incurs the increase of the interference, $p_u(r, \tau, T_u)$ decreases. The expression in Lemma 2 is a closed-form when $\alpha = 4$, since $\zeta_m(T_u) = \frac{\pi}{2} \sqrt{T_u}$. The following lemma provides the DL/UL joint probability which quantifies the performance of the SA-AC scheme with τ . We define the target member which is a member with target distance, r_{tar} . The DL/UL joint probability denotes the probability that a member with r_{tar} from its nearest leader is in the coverage of the leader, i.e., the DL SIR exceeds the DL threshold T_d , and the nearest leader is in the coverage of the member transmitting the UL signal with τ , i.e., the UL SIR exceeds the UL threshold T_u . The performance of the DL/UL joint probability for the target member shows the effect of τ for r_{tar} .

Lemma 3 For τ , the DL/UL joint probability that the DL/UL SIR value for a member with r_{tar} from its nearest leader exceeds T_d and T_u is

$$p_{d,u}(r_{tar}, \tau, T_d, T_u)$$

$$= p_d(r_{tar}, T_d) \tau p_u(r_{tar}, \tau, T_u)$$

$$= \tau e^{-\pi (r_{tar})^2 (\lambda_l \zeta_l(T_d) + \lambda_m p_d(T_d) \tau \zeta_m(T_u))},$$
(16)

where $0 \le \tau \le 1$.

Proof By letting
$$r = r_{tar}$$
 in (7) and (15), $p_{d,u}(r_{tar}, \tau, T_d, T_u)$ is derived.

Since the DL/UL joint probability is affected by T_d , T_u , and r_{tar} , it decreases by increasing them. However, if the SA-AC scheme uses τ^{dyn} as τ , τ^{dyn} increases as T_d increases. Thus, the DL/UL joint probability for the SA-AC scheme is sensitive to T_d . If $\alpha=4$, the closed-form expression of Theorem 3 is derived from the closed-from expressions in Lemma 1 and Lemma 2. From Lemma 3, we expect that τ varies according to r_{tar} . Since $p_{d,u}(r_{tar},\tau,T_d,T_u)$ has the global extreme values for τ , the optimal τ can be derived (see Appendix A.3). We now derive the optimal TP that maximizes $p_{d,u}(r_{tar},\tau,T_d,T_u)$.

Theorem 3 For T_d and T_u , the optimal TP that maximizes the DL/UL joint probability of a target member with r_{tar} is

$$\tau^* = \min\left(\frac{1}{\pi(r_{tar})^2 \lambda_m p_d(T_d) \zeta_m(T_u)}, 1\right). \tag{17}$$

where $0 \le \tau^* \le 1$.

Proof The maximum DL/UL joint probability is found by

$$\frac{\partial p_{d,u}\left(r_{tar},\tau,T_{d},T_{u}\right)}{\partial \tau}=0. \tag{18}$$

Thus the optimum τ is derived.

$$\tau^* = \frac{1}{\pi (r_{tar})^2 \lambda_m p_d(T_d) \zeta_m(T_u)}.$$
 (19)

Since
$$\tau^* \leq 1$$
 in a slot, $\tau^* = 1$ when $r_{tar} < \sqrt{1/\pi \lambda_m p_d(T_d) \zeta_m(T_u)}$.

From Theorem 3, we derive τ^* to maximize the DL/UL joint probability for \bar{r}_{max} . \bar{r}_{max} is the average maximum distance between a leader and its covered member. In general, since r_{tar} has to be estimated by the target member and reported to the leader, \bar{r}_{max} is hard to be known. However, if we assume that λ_m , T_d , and T_u are given, τ^* can be obtained by the average number of the covered members as in the dynamic TP. The average number of the covered members can be known by using the number of associated members. Since the number of the covered members within the distance r_{tar} is $\pi(r_{tar})^2 \lambda_m p_d(T_d)$, $N_m =$ $\pi(\bar{r}_{max})^2 \lambda_m p_d(T_d)$. So $\bar{r}_{max} = \sqrt{\frac{\bar{N}_m}{\pi \lambda_m p_d(T_d)}} = \sqrt{\frac{1}{\pi \lambda_l}}$ by plugging \bar{N}_m in Theorem 1. Thus, $\tau^* = \min\left(\frac{1}{\bar{N}_m \zeta_m(T_u)}, 1\right)$. It implies that τ^* for the average number of the covered members is the same as τ^* for \bar{r}_{max} in the proposed SA-AC scheme.

We now derive the average achievable rate to measure the spectral efficiency performance of the SA-AC schemes. The average achievable rate for a target member

with r_{tar} and the average achievable rate of the members are obtained.

Lemma 4 For T_d and τ , the average achievable rate for a target member with r_{tar} is

$$E_{ar}(r_{tar}, \tau) = \tau \mathbb{E}_{SIR_{d,u}(r_{tar})} \left[ln \left[1 + SIR_{d,u}(r_{tar}) \right] | r_{tar} \right],$$
(20)

where

$$\mathbb{E}_{SIR_{d,u(r_{tar})}} \left[ln \left[1 + SIR_{d,u}(r_{tar}) \right] | r_{tar} \right]$$

$$= \int_0^\infty \exp\left(-\pi \lambda_m p_d(T_d) \tau(r_{tar})^2 (e^t - 1)^{2/\alpha} \right)$$

$$\int_0^\infty \frac{1}{1 + \left(u_{ar,r_{tar}} \right)^{\alpha/2}} du_{ar,r_{tar}} dt.$$
(21)

Furthermore, the average achievable rate of the covered members is

$$E_{ar}(\tau) = \tau \int_0^\infty \frac{1}{1 + \frac{\pi}{2} \frac{\lambda_m p_d(T_d)^2 \tau}{\lambda_l} \left(e^t - 1\right)^{1/2}} dt.$$

$$\approx \frac{\lambda_l}{\lambda_m p_d(T_d)^2}.$$
(22)

Proof See Appendix A.4.
$$\Box$$

Both metrics are computed by averaging the UL coverage probability when the transmission of a covered member is governed by τ . The average achievable rate of a covered member is derived from the average achievable rate for a target member. Both metrics are not closed-form

expressions, and the numerical integration is required to compute them.

5 Numerical results

In this section, we show the numerical results for the analytical models and simulation results with DL threshold T_d and UL threshold T_u . We use the path-loss exponent $\alpha = 4$, the transmit power of a leader $P_l = 1$ W, and the transmit power of a member $P_m = 1 \text{ W [20]}$. The transmission of both the leaders and the members in a single channel is done in multiple time slots. The leader intensity $\lambda_l = 3 \text{ leaders/km}^2$ and the member intensity $\lambda_m = 20$ members/km² [29]. Both the leaders and the members are distributed with their intensity in a 25-km² area. And we focus on the sample area in a 1-km² area to evaluate the performance. Thus, the numerical results within the sample area are obtained in the simulation. In DL, each member computes the DL SIR and checks the DL SIR if it exceeds the DL threshold T_d . If the DL SIR exceeds the DL threshold T_d , the member becomes the covered member. In UL, the covered members transmit their signals according to the TP which is determined by the traditional SA-AC schemes and the proposed SA-AC scheme. The leader computes the UL SIR for each of the covered members and checks if the UL SIR exceeds the UL threshold T_u . if the UL SIR exceeds the UL threshold T_u , the DL/UL of the covered member is successful. We compare the performance of the SA-AC scheme with τ^* to those of the SA-AC schemes with τ^{dyn} as in Theorem 1 and τ^{fix} as in [27].

In Fig. 2, we verify the proposed SA-AC scheme by comparing the performance of the analysis and the simulation.

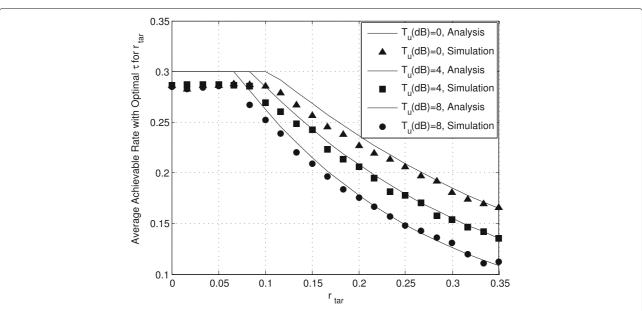


Fig. 2 Average achievable rate with the optimal TP for varying target distance with different T_d when T_d (dB)=-10 (marker: simulation, line: analysis)

The average achievable rate with the optimal τ is presented by varying r_{tar} . Since the performance is analyzed with the PPP assumption, T_d is set to -10. The difference between the results is caused by the effect of approximation for the distribution of the covered members. The distribution of the covered members in the simulation is approximated by PPP in the analysis. The lower T_d is selected, the closer the distribution of the covered members in the simulation approaches that of PPP. The exact distribution is not yet discovered [30]. The difference can be explained by the dependence between the covered members and their leaders. The dependence inherits from T_d . Since the locations of the covered members are jointly changed by T_d and the locations of the leaders, they are not exact PPP. Since the SIR of the DL transmission is difficult to exceed T_d as the distance between a covered member and its leader increases, the average maximum distance between a covered member and its leader decreases as T_d increases. So the distance between the covered member and its leader is not independent. The dependence can be relaxed as T_d decreases. Since the DL coverage probability increases as T_d decreases, the area of a coverage area becomes larger and the distribution of the covered members approaches the PPP distribution of the members.

Since r_{tar} is the distance between the target member and its leader, the signal strength of the target member decreases as r_{tar} increases. Since τ is the TP of the covered members in a slot, the number of the covered members participating in the transmission in a slot increases as τ increases. The signals of the covered members except the target member become the interference to the target member. Thus, for the target member with r_{tar} , the SIR of the target member decreases as r_{tar} increases or τ increases. Since the average achievable rate of a target member with r_{tar} decreases as the SIR decreases, it also decreases as r_{tar} increases or τ increases. If r_{tar} decreases or τ decreases, the average achievable rate for a target member with r_{tar} increases. Since the optimal τ decreases by $(r_{tar})^2$ as in Theorem 3, the interference decreases as r_{tar} increases. However, the signals of the target member and its leader become weaker by α and the number of the transmissions decreases as the optimal τ decreases, then the average achievable rate of the covered member with the optimal τ decreases as r_{tar} increases even if the interference decreases as the optimal τ decreases.

In Fig. 3, the performance of the DL/UL joint probabilities for varying τ is shown with different r_{tar} when $T_d(dB) = -10$ and $T_u(dB) = 0$. The number of transmission members increases as τ increases. Since the interference increases as the number of transmission members increases, the SIR for the member with r_{tar} decreases by increasing τ . However, when the SIR for the member with r_{tar} exceeds T_u even if the interference

for the member increases, $p_{d,u}(r_{tar}, \tau, T_d, T_u)$ increases as τ increases. Once the maximum $p_{d,u}(r_{tar}, \tau, T_d, T_u)$ is achieved, the performance starts to decrease by increasing τ . The larger r_{tar} , the smaller value of the maximum $p_{d,u}(r_{tar}, \tau, T_d, T_u)$ is achieved. Since the signal becomes weaker as r_{tar} increases, the SIR of the member with r_{tar} is hard to exceed T_u by increasing the number of the transmission members. Hence, τ has to be decreased to alleviate the interference and satisfy T_u by reducing the number of the transmission members. Thus, τ to maximize $p_{d,u}(r_{tar}, \tau, T_d, T_u)$ decreases by increasing r_{tar} .

In Fig. 4, the performance of the optimal τ is shown for varying r_{tar} with different T_u when $T_d(dB) = -10$. When the signal strength of the member with r_{tar} is enough to exceed T_u even if all the covered members transmit their data frames, the optimal τ for r_{tar} is one. The number of the inner members decreases and the signal strength increases as r_{tar} decreases. So the optimal τ for r_{tar} increases until one as r_{tar} decreases. However, the optimal τ maximizing $p_{d,u}(r_{tar},\tau,T_d,T_u)$ decreases by increasing r_{tar} as shown in Fig. 3. The smaller T_u , the larger optimal τ is derived. Since the SIR for the member with r_{tar} becomes easy to exceed T_u as T_u decreases, the optimal τ for r_{tar} increases until the increase of the interference is tolerable to the member with r_{tar} .

In Fig. 5, the performance of the DL/UL joint probabilities is shown for varying r_{tar} with the different SA-AC schemes when $T_d(dB) = -10$ and $T_u(dB) = 0$. We use the dynamic τ determined by (9) and (10), $\tau^{dyn} \approx 0.16$. Since the SIR of the member with r_{tar} and the optimal τ to maximize $p_{d,u}(r_{tar})$ decreases as r_{tar} increases, the performance of the SA-AC scheme with the optimal τ decreases as r_{tar} increases. The performance of the SA-AC schemes with dynamic τ and fixed τ decreases as r_{tar} increases. In the schemes with dynamic τ and fixed τ , the values of τ do not vary for r_{tar} , and the SIR of the member with r_{tar} may not exceed T_u as r_{tar} increases. When r_{tar} is smaller than 0.1, the performance of the optimal τ is the same as the case with K of 1. We can expect that the members within r_{tar} satisfies T_u even if all the members try to transmit data with $\tau=1$. However, the DL/UL joint probability of the optimal τ is larger than that of the case with K of 1 when $r_{tar} = 0.15$, since the members with r_{tar} are strongly interfered by the inner members. The performance of the SA-AC scheme using the optimal auapproaches that of the SA-AC scheme using dynamic τ if r_{tar} exceeds 0.1, since dynamic τ is almost the same as the optimal τ for \bar{r}_{max} . The optimal τ is adaptively determined by r_{tar} while τ of the other schemes is not relevant to r_{tar} , thus $p_{d,u}(r_{tar}, \tau, T_d, T_u)$ is always maximized. Conversely, we can induce the distance whose DL/UL joint probability is maximized by τ of the other schemes.

In Fig. 6, the performance of the average achievable rate of the target member is shown for varying r_{tar} with

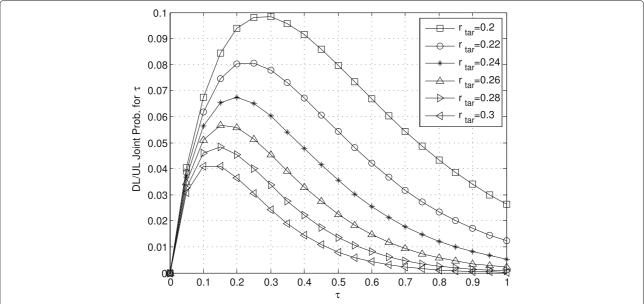


Fig. 3 DL/UL joint probability for varying TP with different target distance when $T_d(dB) = -10$ and $T_u(dB) = 0$. It is shown that there are optimal TPs to maximize the DL/UL joint probability

the different SA-AC schemes when $T_d(dB)=-10$ and $T_u(dB)=0$. We use the dynamic τ as in Fig. 5. Since the signal strength of the target member decreases and the optimal τ decreases to reduce the interference to the target member as r_{tar} increases. So the SIR and the DL/UL joint probability for the member with r_{tar} and the optimal τ decrease as r_{tar} increases. In the SA-AC schemes with dynamic τ and fixed τ , since the interference does

not be alleviated by their TP, the performance of them decreases as r_{tar} increases. However, the optimal τ case maintains the maximum performance. The optimal τ for r_{tar} decreases so that the SIR for the target member with r_{tar} satisfies T_u . However, since the case with K of 1 case among the fixed τ is not adaptive to r_{tar} , the lowest performance is shown when r_{tar} is larger than 0.2. The results show that the decrease of the DL/UL joint probability

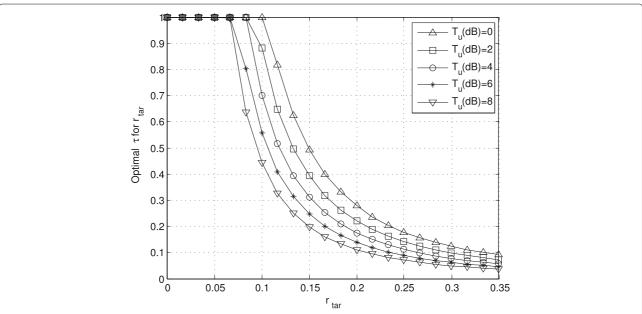


Fig. 4 Optimal TP for varying target distance with different T_d when T_d (dB)=-10. Since the influence of the interference is small in short target distance, the optimal TP is shown to be higher in shorter target distance

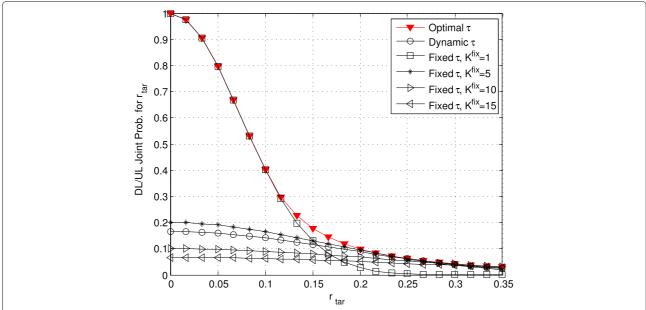


Fig. 5 DL/UL joint probability for varying target distance with the different SA-AC schemes when $T_d(dB) = -10$ and $T_u(dB) = 0$. The proposed SA-AC scheme with optimal TP is shown to be the best by reflecting the spatial effect

induces the degradation of the average achievable rate for r_{tar} .

In Fig. 7, the performance of the DL/UL joint probability for the optimal τ and the performance of the UL probability to maximize the achievable rate UL $_{max}$ τ [20] are shown when $r_{tar}=0.15$ is shown. Since the DL coverage probability decreases as T_d increases, the

number of the covered members decreases. Hence, the interference becomes weaker as T_d decreases, and the SIR of the target member increases as T_d increases. So the DL/UL joint probability increases as T_d increases. Once $p_{d,u}(r_{tar},\tau,T_d,T_u)$ is achieved, the performance starts to degrade by increasing T_d . For r_{tar} , the DL coverage probability of the target member decreases and the optimal

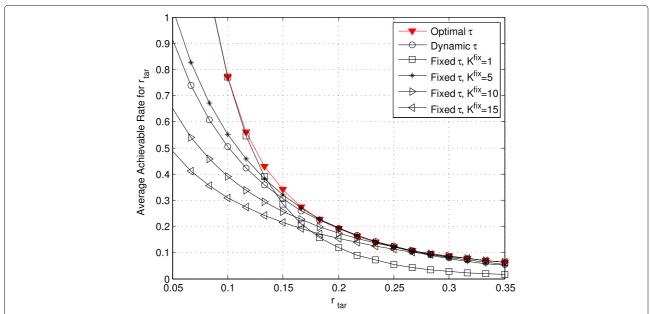
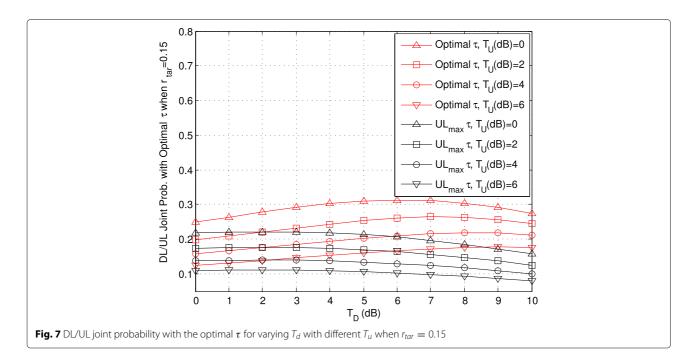


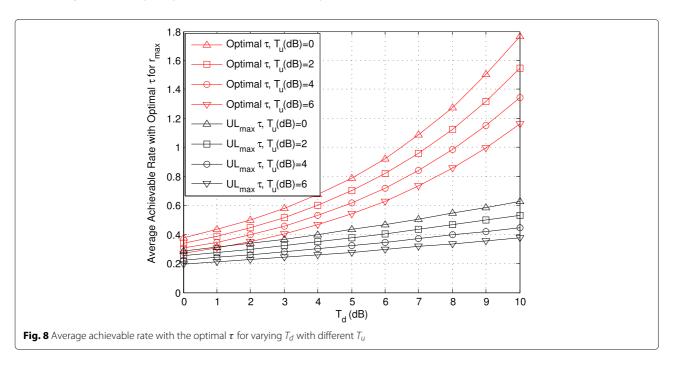
Fig. 6 Average achievable rate for varying target distance with the different SA-AC schemes when $T_d(dB) = -10$ and $T_u(dB) = 0$. As in Fig. 5, the proposed SA-AC scheme with the optimal TP is shown to be the best



au increases as T_d increases. In addition, the DL coverage probability for a distance between a covered member and its leader increases as the distance decreases for T_d . Thus, the optimal au of the covered member with smaller distance from the associated leader of the target member becomes larger. However, since the UL $_{max}$ au is not designed to support group communications and does not consider the number of covered members by the DL threshold, i.e., UL $_{max}$ au is determined to maximize the UL coverage probability only for the member intensity,

the UL_{max} τ is not changed as T_d increases. So the number of transmissions of the covered members does not increase while the Interference decreases as the number of the covered members decreases. The UL SIR of the target member is hard to exceed T_u after the maximum DL/UL joint probability. Since the UL coverage probability decreases as T_u increases, the DL/UL joint probability decreases.

In Fig. 8, the performance of the average achievable rate with the optimal τ and the UL_{max} τ for \bar{r}_{max} is shown



for varying T_d with different T_u . The optimal τ for \bar{r}_{max} varies according to T_d and T_u . Since the increase of T_d reduces the number of the covered members, the interference for a covered member decreases as T_d increases. And the signal strength of a covered member increases as the distance between the covered member and its leader decreases. Then the SIR of the member increases and it is sufficient to exceed T_u . Therefore, the performance of the average achievable rate increases by increasing T_d . The lower T_u , the larger the UL coverage probability becomes by increasing the optimal τ . Thus, the average achievable rate with the optimal τ is maximized by increasing T_d when $T_u(dB) = 0$. However, since the $UL_{max} \tau$ is not changed for the coverage of the leader, it is not possible for the covered members to increase the numbers of transmissions. Although the SIR of the covered member increases by the increase of the signal strength and the decrease of the interference, the $UL_{max} \tau$ might not be sufficient to increase the number of successful covered members in DL/UL.

6 Conclusions

In this paper, we have proposed an SA-AC scheme and developed an analytical model of the SA-AC schemes for group communications network. The proposed analytical model is affected by the intensity of leaders, the intensity of members, and the thresholds for DL and UL which are the important factors of service quality. The proposed analytical model presents the optimal TP to maximize the DL/UL joint probability at a target distance. Since the importance of the DL/UL joint probability at a target distance varies with the type of service, the optimal TP is carefully determined for providing services efficiently. The proposed analytical model has been validated via simulations, and the performance of the SA-AC schemes has been demonstrated. For group communications, the DL/UL joint probability at a target distance can be maximized by the proposed SA-AC scheme, which is superior to other schemes for the target distance. As a result, the importance of considering DL/UL joint probability in the group communication network has been proven.

Appendix

A.1 Proof of Lemma 1

For the DL threshold T_d , the DL coverage probability that a member with the distance r from its nearest leader is

$$p_{d}(r, T_{d}) = \mathbb{P}\left[\operatorname{SIR}_{d}(r) > T_{d}|r\right]$$

$$= \mathbb{P}\left[h_{l,m} > P_{l}^{-1}r^{\alpha}I_{l}T_{d}|r\right]$$

$$= \mathbb{E}\left[\exp\left(P_{l}^{-1}r^{\alpha}I_{l}T_{d}\right)|r\right]$$

$$= \mathcal{L}_{I_{l}}\left(P_{l}^{-1}r^{\alpha}T_{d}\right), \tag{23}$$

where $\mathcal{L}_{I_l}(s)$ is the Laplace transform of the aggregate interference received at the leader. $\mathcal{L}_{I_l}(s)$ can be expressed as

$$\mathcal{L}_{I_{l}}(s) = \mathbb{E}_{I_{l}}\left[e^{-sI_{l}}\right] \\
= \mathbb{E}_{\Phi_{l},h_{l_{i},m}}\left[e^{-s\left(\sum_{l_{i}\in\Phi_{l}\setminus l}P_{l}h_{l_{i},m}r_{d,i}^{-\alpha}\right)}\right] \\
= \mathbb{E}_{\Phi_{l}}\left[\prod_{l_{i}\in\Phi_{l}\setminus l}\mathbb{E}_{h_{l_{i},m}}\left[e^{-h_{l_{i},m}\left(sP_{l}r_{d,i}^{-\alpha}\right)}\right]\right] \\
\stackrel{(a_{1})}{=}e^{-2\pi\lambda_{l}\int_{r}^{\infty}\left(1-\mathbb{E}_{h_{l_{i},m}}\left[e^{-h_{l_{i},m}\left(sP_{l}v_{l}^{-\alpha}\right)}\right]\right)\nu_{l}d\nu_{l}} \\
= e^{-2\pi\lambda_{l}\int_{r}^{\infty}\left(1-\frac{1}{1+sP_{l}v_{l}^{-\alpha}}\right)\nu_{l}d\nu_{l}} \\
\stackrel{(a_{2})}{=}e^{-2\pi\lambda_{l}\int_{r}^{\infty}\left(\frac{1}{1+\left(v_{l}/\left(r\left(T_{d}\right)^{1/\alpha}\right)\right)^{\alpha}}\right)\nu_{l}d\nu_{l}} \\
\stackrel{(a_{2})}{=}e^{-\pi r^{2}\lambda_{l}T_{d}^{2/\alpha}}\int_{T_{d}^{-2/\alpha}}^{\infty}\frac{1}{1+\left(u_{l}\right)^{\alpha/2}}du_{l}} \\
\stackrel{(a_{3})}{=}e^{-\pi r^{2}\lambda_{l}\zeta_{l}\left(T_{d}\right)}, \tag{24}$$

where (a_1) is derived from the probability generating functional (PGFL) of PPP [31], (a_2) uses $s=P_l^{-1}r^\alpha T_d$, (a_3) is derived by the change of the variable $u_l=\left(v_l/r(T_d)^{1/\alpha}\right)^2$, and (a_4) utilizes the change of the variable $\zeta_l(T_d)=T_d^{2/\alpha}\int_{T_d^{-2/\alpha}}^\infty \frac{1}{1+(u_l)^{\alpha/2}}\mathrm{d}u_l$. Thus, plugging (24) into (23), we obtain (8). And combining (23) with (7), we obtain the DL coverage probability that a member is in the coverage of its nearest leader given by

$$p_d(T_d) = \int_0^\infty p_d(r, T_d) f_R(r) dr.$$
 (25)

Lemma 1 and its derivation corresponds to the interference limited network case as in [4].

A.2 Proof of Lemma 2

For the UL threshold T_u and τ , the UL coverage probability that the SIR value for the transmission of the covered member with the distance r from its leader exceeds T_u is given by

$$p_{u}(r, \tau, T_{u}) = \mathbb{P}\left[\operatorname{SIR}_{d,u}(r, \tau) > T_{u}|r\right]$$

$$= \mathbb{P}\left[h_{m_{c},l} > P_{m}^{-1}r^{\alpha}I_{m_{c}}T_{u}|r\right]$$

$$= \mathbb{E}\left[\exp\left(P_{m}^{-1}r^{\alpha}I_{m_{c}}T_{u}\right)|r\right]$$

$$= \mathcal{L}_{I_{m_{c}}}\left(P_{m}^{-1}r^{\alpha}T_{u}\right)$$
(26)

where $\mathcal{L}_{I_{m_c}}(s)$ denotes the Laplace transform of the aggregate interference received at the leader. $\mathcal{L}_{I_{m_c}}(s)$ can be found in a similar way to (24). Thus,

$$\mathcal{L}_{I_{m_c}}(s)$$

$$= \mathbb{E}_{I_{m_c}} \left[e^{-sI_{m_c}} \right]$$

$$= \mathbb{E}_{\Phi_{m_c}, h_{m_{c,i},l}} \left[e^{-s\left(\sum_{m_{c,i} \in \Phi_{m_c} \setminus m_c} P_m h_{m_{c,i},l} r_{du,i}^{-\alpha}\right)} \right]$$

$$= \mathbb{E}_{\Phi_{m_c}} \left[\prod_{m_{c,i} \in \Phi_{m_c} \setminus m_c} \mathbb{E}_{h_{m_{c,i},l}} \left[e^{-h_{m_{c,i},l} \left(s P_m r_{du,i}^{-\alpha}\right)} \right] \right]$$

$$= \exp\left(-2\pi \lambda_m p_d(T_d) \tau$$

$$\int_0^\infty \left(1 - \mathbb{E}_{h_{m_{c,i},l}} \left[e^{-h_{m_{c,i},l} \left(s P_m v_m^{-\alpha}\right)} \right] \right) \nu_m d\nu_m \right)$$

$$= e^{-2\pi \lambda_m p_d(T_d) \tau} \int_0^\infty \left(1 - \frac{1}{1 + s P_m v_m^{-\alpha}} \right] \right) \nu_m d\nu_m$$

$$\stackrel{(a_5)}{=} e^{-2\pi \lambda_m p_d(T_d) \tau} \int_0^\infty \left(\frac{1}{1 + \left(v_m / (r(T_u)^{1/\alpha})\right)^{\alpha}} \right) \nu_m d\nu_m$$

$$\stackrel{(a_6)}{=} e^{-\pi r^2 \lambda_m p_d(T_d) \tau} T_u^{2/\alpha} \int_0^\infty \frac{1}{1 + (u_m)^{\alpha/2}} du_m$$

$$\stackrel{(a_7)}{=} e^{-\pi r^2 \lambda_m p_d(T_d) \tau} T_u^{2/\alpha} \int_0^\infty \frac{1}{1 + (u_m)^{\alpha/2}} du_m$$

where (a_5) is found from $s = P_m^{-1} r^{\alpha} T_u$, (a_6) comes from the change of variable $u_m = (v_m/r(T_u)^{1/\alpha})^2$, and (a_7) is derived from the change of variables

$$\zeta_m(T_u) = T_u^{2/\alpha} \int_0^\infty \frac{1}{1 + (u_m)^{\alpha/2}} du_m = T_u^{2/\alpha} \frac{2\pi/\alpha}{\sin(2\pi/\alpha)}.$$
(28)

A.3 Proof of Optimal au

Let $f(\tau)$ be $p_{d,u}(r_{tar}, \tau, T_d, T_u)$. By the extreme value theorem, if $f(\tau)$ is continuous in the interval of $\tau \in [0, 1]$, then at some point of the interval, $f(\tau)$ attains a global maximum and a global minimum [32–34]. Then the point of attainment is either 1) a point where $f'(\tau)$ does not exist, 2) a point where $f'(\tau) = 0$, and 3) a point at one end of the interval [33]. $f(\tau)$ can be expressed as

$$f(\tau) = \tau \exp(-a - b\tau),\tag{29}$$

where $a = \pi r_{tar}^2 \lambda_l \zeta_l(T_d)$ and $b = \pi r_{tar}^2 \lambda_m P_l^c(T_d) \zeta_m(T_u)$, $a \ge 0$ and $b \ge 0$. Since $f(\tau)$ is differentiable in $\tau \in [0, 1]$, $f(\tau)$ is continuous and $f'(\tau)$ exists for all $\tau \in [0, 1]$.

$$f'(\tau) = (1 - b\tau) \exp(-a - b\tau). \tag{30}$$

When $\tau = \frac{1}{b}$, $f'(\tau) = 0$. We evaluate $f(\tau)$ at the points where $f'(\tau) = 0$ and at the end points as the candidates for the extreme values. The function value at the point where $f'(\tau) = 0$ is

$$f\left(\frac{1}{b}\right) = \frac{1}{b}\exp(-a - 1),\tag{31}$$

The function values at the end points are

$$f(0) = 0, (32)$$

$$f(1) = \exp(-a - b). \tag{33}$$

Since $f(\tau) \geq 0$, f(0) = 0 is the global minimum of $f(\tau)$. If $0 \leq b < 1$, $\frac{1}{b} > 1$ and $f\left(\frac{1}{b}\right)$ is not defined in the interval of $\tau \in [0,1]$. Thus, f(1) is the global maximum when $0 \leq b < 1$. If $b \geq 1$, then $be^1 \leq e^b$. Thus, $f\left(\frac{1}{b}\right) \geq f(1)$ can be verified as follows.

$$\frac{1}{be^1} \ge \frac{1}{e^b}.\tag{34}$$

Thus, $f\left(\frac{1}{b}\right)$ is the global maximum of $f(\tau)$ when $b \ge 1$. Therefore an optimal τ exists for $f(\tau)$ in the interval of [0,1].

A.4 Proof of Lemma 4

For T_d and τ , the average achievable rate for a target member with r_{tar} is given by

$$E_{ar}(r_{tar}, \tau)$$

$$= \tau \int_{0}^{\infty} \mathbb{P}\left[\ln\left(1 + \frac{P_{m}h_{m_{c},l}r_{tar}^{-\alpha}}{I_{m_{c}}} > t\right)\right] dt$$

$$= \tau \int_{0}^{\infty} \mathbb{P}\left[h_{m_{c},l} > P_{m}^{-1}r_{tar}^{\alpha}I_{m_{c}}(e^{t} - 1)\right] dt$$

$$= \tau \int_{0}^{\infty} \mathbb{E}\left[\exp\left(P_{m}^{-1}r_{tar}^{\alpha}I_{m_{c}}(e^{t} - 1)\right)\right] dt$$

$$= \tau \int_{0}^{\infty} \mathcal{L}_{I_{m_{c}}}\left(P_{m}^{-1}r_{tar}^{\alpha}(e^{t} - 1)\right) dt. \tag{35}$$

Similar to Appendix A.1 and Appendix A.2, it follows,

$$\mathcal{L}_{I_{m_c}} \left(P_m^{-1} r_{tar}^{\alpha}(e^t - 1) \right) \\
= e^{-2\pi \lambda_m p_d (T_d) \tau} \int_0^{\infty} \left(1 - \frac{1}{1 + (r_{tar}/\nu_m)^{\alpha}(e^t - 1)} \right] \nu_m d\nu_m \\
= e^{-2\pi \lambda_m p_d (T_d) \tau} \int_0^{\infty} \left(\frac{1}{1 + (\nu_m/(r(e^t - 1)^{1/\alpha}))^{\alpha}} \right) \nu_m d\nu_m \\
\stackrel{(a_8)}{=} \exp\left(-\pi \lambda_m p_d (T_d) \tau (r_{tar})^2 (e^t - 1)^{2/\alpha} \right) \\
\int_0^{\infty} \frac{1}{1 + (u_{ar, r_{tar}})^{\alpha/2}} du_{ar, r_{tar}} \right).$$
(36)

where (a_8) uses the change variable $u_{ar,r_{tar}} = \left(\frac{v_m}{r_{tar}(e^t-1)^{1/\alpha}}\right)^2$.

When $\alpha = 4$ and $r_{tar} = r$, the average achievable rate of a covered member is then given by

$$E_{ar}(\tau)$$

$$= \tau \int_{0}^{\infty} \int_{0}^{\infty} \mathcal{L}_{I_{mc}} \left(P_{m}^{-1} r^{\alpha} \left(e^{t} - 1 \right) \right) f_{R_{mc}}(r) dt dr.$$

$$\stackrel{(a_{9})}{=} \frac{\pi \lambda_{l} \tau}{p_{d}(T_{d})} \int_{0}^{\infty} \frac{2p_{d}(T_{d})}{2\pi \lambda_{l} + \pi^{2} \lambda_{m} p_{d}(T_{d})^{2} \tau (e^{t} - 1)^{1/2}} dt.$$

$$= \tau \int_{0}^{\infty} \frac{1}{1 + \frac{\pi}{2} \frac{\lambda_{m} p_{d}(T_{d})^{2} \tau}{\lambda_{l}} (e^{t} - 1)^{1/2}} dt.$$

$$\stackrel{(a_{10})}{\approx} \tau \int_{0}^{\infty} \frac{1}{\frac{\pi}{2} \frac{\lambda_{m} p_{d}(T_{d})^{2} \tau}{\lambda_{l}} (e^{t} - 1)^{1/2}} dt.$$

$$\stackrel{(a_{11})}{=} \tau \int_{0}^{\infty} \frac{1}{u^{2} + C^{2}} du = \tau \left[\frac{1}{C} \arctan \left(\frac{u}{C} \right) \right]_{0}^{\infty}$$

$$= \tau \frac{\pi}{2C} = \frac{\lambda_{l}}{\lambda_{m} p_{d}(T_{d})^{2}}$$

$$(37)$$

where (a_9) utilizes the change of the variable $z=r^2$ and (9) then change the order of integrals. (a_{10}) comes from the approximation of $1/\left(1+a\sqrt{e^t+1}\right)$ to $1/\left(a\sqrt{e^t+1}\right)$ where a>0. Also, (a_{11}) stems from the change of variables as $C=\frac{\pi\lambda_m p_d(T_d)^2\tau}{2\lambda_l}$ and $u=C\sqrt{e^t-1}$.

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Authors' contributions

ML's contribution is writing the paper and conducting performance analysis and simulations. YK's contribution is partly writing the paper and conducting simulations. T-JL's contribution is writing and revising the paper and guiding the direction and organization of the paper. All authors read and approved the final manuscript.

Competing interests

The authors declare that they have no competing interests.

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