Understanding the Impact of Interference on Collaborative Relays

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Abstract—Collaborative relays achieve the benefits of spatial diversity without requiring physical antenna arrays at end devices. While many studies have demonstrated its effectiveness in an isolated source-destination system, applying cooperative relays to large-scale wireless networks remains challenging. We show that a system with cooperative relays can be penalized by the extra level of interference it produces. By examining the inter-dependency between interference management and cooperative relay strategies, we model the penalty by an increase in spectrum resource usage and translate it into a penalty in single-link throughput. This penalty serves as a reference for cooperative relay design in isolated scenarios. We then present two practical spectrum allocation mechanisms for collaborative relay systems with different fairness and utilization tradeoffs. Simulation results confirm our analytically findings.

Index Terms—Wireless network, collaborative relay, interference management, resource allocation.

I. INTRODUCTION

W IRELESS networks suffer from frequent and unpredictable packet losses because of channel fading. Many wireless systems use multiple antenna arrays (MIMO) which exploit spatial diversity to mitigate channel fading and extend transmission range. Most portable wireless devices, however, can carry 1-2 antennas, significantly limiting the effectiveness of MIMO mechanisms.

Recent proposals on collaborative relays [1]–[3] address this limitation by node collaborating with each other. Devices equipped with omnidirectional antennas can intersect a neighboring transmitter's signal and relay it to the designated receiver. The receiver combines multiple signal streams from both source and relay nodes to recover signals with higher reliability. By collaboratively relaying signals, nodes form a *virtual* antenna array, achieving the same benefits of a MIMO system with only one antenna per device.

There have been significant efforts on understanding and improving the benefits of collaborative relays in small (isolated) wireless networks. In a simple single source-destination-relay scenario, existing contributions span from analysis of channel capacity [4], [5] and diversity-multiplexing tradeoff [2], [6], to algorithm design for power allocation [7], [8], relay selection [9] and link retransmission [10]. Others extend the results to a more complex scenario with multiple sourcedestination-relay nodes, including capacity analysis [11], resource allocation [12], power allocation [8] and the use of group-nulling to increase array gain [13]. Overall, these studies have demonstrated the effectiveness of collaborative relays in small-scale isolated wireless networks.

Deploying collaborative relays in large-sale wireless networks, however, faces many challenges [9], [14], [15]. In addition to requiring efficient protocols for node collaboration, one critical challenge is how to address the impact of extra interference produced by signal relaying. When interference is a dominating limiting factor, increase in interference level translates into degraded throughput and higher packet losses.

In this paper, we show both analytically and experimentally that the use of collaborative relays in large wireless networks is penalized by the extra level of interference. We characterize the penalty by an increase in the amount of channel resource consumption, and translate it into a throughput penalty in an isolated source-destination-relay scenario. This throughput penalty provides a reference for designing collaborative relay strategies. In particular, we show that a large-scale wireless network can benefit from the use of collaborative relays only if the gain of relaying in isolated scenarios is higher than a critical ratio T > 1 rather than T = 1.

In addition to identifying and characterizing the interference penalty, we also propose channel allocation mechanisms to mitigate the impact of interference. We propose to assign orthogonal channels to conflicting transmission links. By assigning different priorities to direct and relay links, we design two approaches that tradeoff between fairness and system throughput. We perform extensive experiments to verify our analytical findings and evaluate the proposed channel allocation mechanisms.

The rest of the paper is organized as follows. We start from a brief overview of collaborative relays in Section II.

We then model the network-wide interference in Section III. In Section IV, we use this model to define a conflict-graph based resource allocation problem, and mitigate interference by assigning different channels to conflicting transmissions. Next, we examine the asymptotic interference condition and analytically derive the penalty of relay interference in Section V. In Section VI we conduct simulations to evaluate the performance of the proposed algorithms and validate the theoretical findings. Finally, we discuss design implications and future directions in Section VII, and conclude in Section VIII.

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Fig. 1. Cooperative Relays.

II. BACKGROUND ON COLLABORATIVE RELAY

Collaborative relays form a virtual multiple antenna system by nodes relaying signals for one another. Existing studies have proposed several relay strategies with different performance and complexity tradeoff. In this section, we briefly describe a representative set of relay strategies.

A. Collaborative Relay Strategies

Figure 1 shows a basic collaborative relay system. Using omnidirectional antennas, a transmitter S transmits a signal to its designated receiver D. A collaborative neighbor R intersects the signal and forwards it to D. D combines two signal streams, $S \Rightarrow D$ and $R \Rightarrow D$ into a single stream which has higher resistance to channel fading and noise, hence, higher probability of being successful decoded. In general, there are two types of collaborative relays [2].

Amplify and forward (AF). R receives signals from S, amplifies and forwards them to D without demodulation or decoding. As a result, it also amplifies the noise associated with link $S \Rightarrow R$. When the time gap between direct and relay paths is small, D can use a RAKE receiver to combine both signal streams [2], [16]. Hence the system behaves like a single link with one transmit antenna and two receive antennas. We refer to this strategy as AF-RAKE. Note that AF-RAKE requires a full-duplex radio at R that transmits and receives from the same channel concurrently. A low-cost alternative is AF-FDD where R receives from S on one channel and forwards the signal on a different channel, while D receives from both channels and combines the signals. This requires Rand D to use two half-duplex radios and two channels. When only a single half-duplex radio is available at each node, the system can multiplex $S \Rightarrow D$ and $R \Rightarrow D$ into two time slots, referring to as AF-TDD.

Decode and forward (DF). R decodes the signal from S, re-encodes it with sophisticated coding techniques [2], [5] and forwards the re-generated signal to D. Since R has to decode and re-encode the signal, $S \Rightarrow D$ and $R \Rightarrow D$ are carried out in two time slots. Note that this operation only requires a half-duplex radio at R and D and one channel.

In this paper, we focus on AF strategies to examine the system performance of collaborative relay, rather than proposing any new coding and transmission technologies. However, our work can be easily extended to include other relay strategies.

B. Capacity Abstractions

Next, we utilize information theoretical capacity to characterize per-link transmission efficiency of collaborative relays. The capacity of a transmission represents the maximum amount of information per second per Hertz that can be sent successfully. It provides a good approximation of link efficiency in real wireless systems without delving into complex coding, detecting and decoding procedures. In this paper, we use ergodic and outage capacities to characterize a wireless fading channel, and use throughput which is the product of the channel capacities and spectrum bandwidth to characterize transmission quality. Next, we describe these two capacities in details.

Ergodic Capacity represents the expectation of the instantaneous channel capacity, averaged across channel fading. It captures the average packet throughput for delay non-sensitive applications. We can approach ergodic capacity by adapting transmission settings (*e.g.* power, modulation, coding rate) to channel variations.

Outage Capacity measures the maximum data rate supported with probability of $1 - \zeta$ assuming a fixed transmission rate. ζ defines the outage probability or the maximum tolerable packet loss rate. An outage occurs if the transmission rate is higher than the instant channel capacity, (*e.g.*, no packet can be sent successfully). Different from ergodic capacity, outage capacity captures the maximum supported data rate for delay sensitive applications such as VoIP and video streaming.

In general, there are no close-form expressions of ergodic and outage capacities. The work in [2] provides simplified approximations of both capacities assuming each node transmits at a constant power and experiences independent frequency flat fading. Let γ represent the transmit signal-to-noise ratio and W represent the spectrum bandwidth of each channel, we derive the capacities of direct transmission and AF-RAKE as follows:

Ergodic Capacity:

$$C_{E,DT} = E_{h_{SD}(t)} \log \left(1 + |h_{SD}(t)|^2 \gamma\right)$$

$$C_{E,AF-RAKE} = E_h \log \left(1 + |h_{SD}(t)|^2 \gamma + f\left(|h_{SR}(t)|^2 \gamma, |h_{RD}(t)|^2 \gamma\right)\right) \quad (1)$$

where

$$f(x,y) := \frac{xy}{x+y+1}$$

Outage Capacity at an outage probability of ζ :

$$C_{O,DT} = \log(\gamma \zeta \sigma_{s,d}^2 + 1)$$
$$C_{O,AF-RAKE} = \log(\gamma \sqrt{\zeta \phi} + 1)$$
(2)

where $\frac{1}{\phi} := \frac{1}{2\sigma_{s,d}^2} \frac{\sigma_{s,r}^2 + \sigma_{r,d}^2}{\sigma_{s,r}^2 \sigma_{r,d}^2}$ and $\sigma_{i,j}^2 := E(|h_{ij}(t)|^2)$. $h_{ij}(t)$ represents the channel response from node *i* to *j* at time *t* that captures the effects of path-loss, shadowing and fading.

Table I provides a summary of various relay strategies. Note that the capacities of AF-TDD are half of those of AF-RAKE since AF-TDD splits the transmissions into two time slots. The throughputs of AF-FDD are the same as those of AF-RAKE assuming that in AF-RAKE, S, R and D all operate at a single channel of bandwidth W, while in AF-FDD, they operate at two channels, each of bandwidth W. Therefore, per-band throughput of AF-FDD is half of that of AF-RAKE. Figure 2 plots the throughputs of different



Fig. 2. Throughputs of different cooperative strategies for W = 1.

strategies as a function of γ , assuming W = 1 and all channels follow Rayleigh fading with $\sigma_{i,j}^2 = 1$. We see that AF-RAKE and AF-FDD significantly outperform direct transmission; and AF-TDD is slightly beneficial in terms of outage throughput for small signal-to-noise ratio. Overall, AF schemes achieve more performance improvement in terms of outage capacity. This is because transmitting via two independent paths can significantly improve packet delivery rate. For more details about ergodic and outage capacities, please see [17, Ch. 5] and [2].

III. MODELING COLLABORATIVE RELAY IN AD HOC NETWORKS

While results in Figure 2 demonstrate the effectiveness of collaborative relay in a network of 3 nodes, deploying collaborative relays in a large network must consider the impact of elevated interference from relaying. In this section, we present an analytical model on network-wide interference in collaborative relay enabled large scale networks. We start from describing two general interference models, and extend them to collaborative relay based networks.

A. General Interference Model

We consider a large-scale network with static nodes and single-hop transmissions. We use two receiver-driven interference models, the *Physical Model* and the *Protocol Model*, to characterize radio transmissions in the presence of interference. The physical model was introduced in [18] to integrate the impact of channel fading while the protocol model is a simplified version [19], [20].

Physical Model. Let S denote the set of wireless nodes that communicate on the same channel¹. when node i' communicates to node i, the received signal-to-interference-noise ratio at node i is defined as

$$SINR_{i}(t) = \frac{P_{i'}|h_{ii'}(t)|^{2}d_{ii'}^{-\alpha}}{\sigma^{2} + I_{i}(t)},$$

$$I_{i}(t) = \sum_{j \in \mathbb{S}, j \neq i'} P_{j}|h_{ij}(t)|^{2}d_{ij}^{-\alpha}.$$
 (3)

where P_j is the transmit power at neighboring node j, σ^2 represents the thermal noise variance, $h_{ij}(t)$ represents the smallscale channel fading from transmitter j to receiver i at time t, and d_{ij} is the distance between i and j. The large-scale fading (path loss) is approximated by $d_{ij}^{-\alpha}$ where $\alpha = 2$ for free space propagation and $\alpha > 2$ for general environments [18], [21]. Each link's transmission can be successfully decoded only if its received SINR is above a given threshold $SINR_0$ for a given transmission rate. Therefore, at each time instance, the total level of tolerable interference at node i, defined by $I_{0,i}$ is

$$I_{0,i}(t) = \frac{P_{i'}|h_{ii'}(t)|^2 d_{ii'}^{-\alpha}}{SINR_0} - \sigma^2.$$
 (4)

Given a set of transmissions in *i*'s neighborhood, only if $I_i(t) \leq I_{0,i}(t)$, *i'* can successfully communicate to *i* at a particular rate determined by $SINR_0$.

Protocol Model. While providing a sophisticated model of interference, the physical model also requires extensive knowledge of channel instance h_{ij} , which is computational-expensive to obtain. A simplified version is to use a no-talk distance r_I . Assuming each user transmits at a constant power P, for each receiver i, if all the transmitters except the node i' are more than $r_I(i)$ distance away, then with very high probability that the overall accumulated interference is less than the tolerable threshold, *i.e.*

$$Prob\{I_i(t) \le I_{0,i}(t)\} > 1 - \epsilon, \tag{5}$$

where $\epsilon \ll 1$ represents tolerable packet loss rate. In another word, $r_I(i)$ is a location-dependent distance-driven *interference-sensitive range* where if any transmitter is within $r_I(i)$ distance from *i*, it could potentially interfere with *i*'s reception. Note that the value of $r_I(i)$ depends on the level of transmit power *P*, the transmit power density, radio propagation condition, transmission frequency, $SINR_0$ and ϵ . Assuming negligible background noise, $r_I(i)$ has a simple lower bound [20]:

$$r_I(i) \ge SINR_0^{\frac{1}{\alpha}} d_{ii'} \tag{6}$$

¹Note that the destination nodes are determined by source nodes and vice versa. Therefore, we identify these two types of nodes.

TABLE I
SUMMARY OF RELAY STRATEGIES

	No. of channels	No. of time slots	Decode at relay	Duplexity	Throughput
AF-RAKE	1	1	No	Full	$W \cdot C_{AF-RAKE}$
AF-TDD	1	2	No	Half	$\frac{1}{2}W \cdot C_{AF-RAKE}$
AF-FDD	2	1	No	$(2\times)$ Half	$W \cdot C_{AF-RAKE}$
DF	1	2	Yes	Half	$W \cdot C_{DF}$ [2]

In order to applying the general interference models to large-scale wireless networks, we make the following assumptions:

- Each link's interference sensitive region is the combination of two nodes. This is because we assume transmissions between any two nodes require handshaking, *e.g.* RTS-CTS-DATA-ACK in IEEE 802.11. Therefore, each node of a link pair becomes both a transmitter and receiver, *i.e.* a transceiver. Any transmitter within their interference-sensitive range could degrade the reception. Therefore, *interference-sensitive region* of a link is the combination of those of both individual nodes. Figure 3(a) shows an example. Overall, any node that falls into the interference-sensitive region of a link will potentially interfere the transmission.
- 2) We only focus on the protocol model because of its simplicity. While the physical model provides accurate modeling of interference, applying it to a large scale system is computational-expensive. Even as a pessimistic binary simplification of the physical model, the protocol model captures the the essence of accumulated interference from neighboring transmitters as well as the impact of radio propagation model, without delving into the details of demodulation, decoding and signal processing.
- 3) We assume nodes transmit at a fixed power. Instead of adjusting transmit power level to reduce interference, we focus on using spectrum redundancy to assign transmissions that could possibly interfere one another with orthogonal non-interfering channels or time slots, while allowing spectrum reuse between links that are wellseparated.

With these assumptions in mind, we now describe the interference model for cooperative relay.

B. Collaborative Relay Interference Model

Physical Model. Let S and \mathbb{R} denote the sets of source nodes and the relay nodes that communicate on the same channel, respectively. Assume no nodes can play two roles, *e.g.* S and \mathbb{R} are disjoint. A transmission from i' to i which is assisted by an amplify-and-forward collaborative relay, l, has a received SINR of

$$SINR_{i}(t) = \frac{P_{i'}|h_{ii'}(t)|^{2}d_{ii'}^{-\alpha} + P_{l}\frac{SINR_{l}(t)}{SINR_{l}(t)+1}}{\sigma^{2} + \sum_{j \in \mathbb{S} \bigcup \mathbb{R}, j \neq i', j \neq l} P_{j}|h_{ij}(t)|^{2}d_{ij}^{-\alpha} + P_{l}\frac{1}{(SINR_{l}(t)+1)}}$$
(7)



Fig. 3. Interference-sensitive regions of (a) a direct link (b) a collaborative relay

where

$$SINR_{l}(t) = \frac{P_{i'}|h_{li'}(t)|^{2}d_{li'}^{-\alpha}}{\sigma^{2} + \sum_{j \in \mathbb{S} \bigcup \mathbb{R}, j \neq i'} P_{j}|h_{lj}(t)|^{2}d_{lj}^{-\alpha}}$$
(8)

Note that for AF, relay node l amplifies its perceived interference to node i. Therefore, it is equally important to prevent interference at relay and destination nodes. At the same time, relay node l extends the interference range of the direct transmission. These observations lead us to re-define the interference-sensitive regions of collaborative relays in the protocol model.

Protocol Model. Following the general protocol model, we approximate the interference-sensitive region of a collaborative relay as the combination of interference regions of source, relay and destination nodes, as shown in Figure 3(b).

Overall, each transmission (direct or relay) has an interference-sensitive region which is the combination of the regions associated with the nodes involved. Any transmitter in the sensitive region will produce non-negligible interference to the receiver and degrade its performance. While prior work focuses on applying power control to reduce interference, we take a different approach to utilize spectrum redundancy. Assuming a multi-channel system, we focus on channel allocations that assigns orthogonal channels to conflicting links while allowing well-separated links to reuse channels. Note that the concept of channels is logical as a channel can be a physical frequency band, a time slot, or a channelization code.

IV. RESOURCE ALLOCATION FOR COLLABORATIVE RELAY

From the interference model in Section III, we see that a collaborative relay has much larger interference-sensitive region, and also produces a higher level of interference to peers. Such increased interference sensitivity leads to a penalty in system performance. In this section, we study the penalty of extra interference by examining channel allocations in collaborative relay networks. In particular, we compare the amount of spectrum resource each link can get with and without using relays. We start with modeling the corresponding resource allocation problem into a conflict-graph based graph-coloring problem, and then propose different allocation mechanisms.

A. Reduction to Conflict Graph and Graph Coloring

We show that the network-wide interference can be modeled by a conflict graph. We introduce a graph G(V, E) with vertices V and undirected edges E [22]. Each vertex $u \in V$ represents a link rather than a single communication node. An edge exists between two link-vertices u and v if they conflict and must use different channels. Assuming symmetric interference patterns, the edges are undirected.

In Figure 4, we illustrate the procedure to build a conflict graph from a network topology.

- 1) **Direct Transmissions:** Each link maps to a vertex in the conflict graph. Any two vertices connect if one falls into the other's interference-sensitive region.
- AF-RAKE: Combine both the direct and relay links into a single vertex because they operate on the same channel. For example, in Figure 4, we combine the node S1, R1, D1 and their associated links into a single vertex (L1).
- AF-FDD: The direct and relay links use different channels and hence map to two vertices. The edges should be constructed similarly as previous two cases.

For simplicity, we assume a network with M channels of equal bandwidth, slow-varying network topology and traffic volume. The resource allocation problem is how to assign channels to links so that they do not conflict. Because of the interference constraints, this problem is known to be NP-complete [23].

B. Allocating Channels for Direct Transmissions

Fig. 4. An Example of Conflict Graph

Existing study converts channel allocation problems into graph coloring problems by mapping each channel to a color [24], [25]. The coloring algorithm is constrained by that if an edge exists between any two vertices, they can not simultaneously use one color. While the general coloring problem is known to be NP-hard [23], existing work shows that heuristic based approaches produce good approximations to the optimal solution. When the number of channels M is adjustable, the algorithms in [24] can be applied to derive the minimum number of channels required for each link to obtain a channel, i.e. to maximize per-channel bandwidth 1/M. When the number of channels is fixed, the algorithms in [25] can be applied to maximize the number of links being assigned with a channel. Distributed algorithms such as local bargaining and rule guided self-adjustment [26], [27] provide good approximations to the graph coloring approaches.

The performance of channel allocation depends on the maximum vertex degree of the conflict graph. The degree of a vertex is the number of vertices which is connected to, *i.e.* the number of conflicting neighbors. The number of channels required to color a conflict graph is upper bounded by the maximum vertex degree +1 [22]. Compared to direct transmissions, collaborative relays result in conflict graphs with higher vertex degree and hence require more channels to achieve conflict-free transmissions. In other words, when the number of channels is small, some links might be disabled due to lack of channels.

C. Allocating channels to collaborative relays

We extend the algorithms in [25] to distribute channels among transmission links. Using AF-RAKE, we assign the same channel to both the direct link $S \Rightarrow D$ and the relay link $R \Rightarrow D$. Using AF-FDD, we assign two different channels to these links. However, when the number of channels is small and hence some links are blocked, we need to choose the links intelligently to maximize system utility.

Next, we present two different allocation strategies that assign different priorities to direct and relay links.



Collaborative Allocation (CA). This allocation scheme prioritizes direct links over relay links by performing allocation sequentially starting from direct links. The system first allows direct links to negotiate for channel usage using schemes in [25], [26]. Only direct links with channel allocated can set up their relay links. Using AF-RAKE, a relay is set up only if the channel assigned to the direct link is available at the relay link, *i.e.* it does not conflict with neighboring links with channel assigned. Using AF-FDD, a relay is set up only if a channel different from that of the direct link is available. Overall, using such sequential assignment, CA scheme is fairness-driven, *i.e.* it attempts to minimize the impact of relay links on direct links. Although minimizing the probability of blocking direct transmissions, this strategy may not be the optimal strategy for maximizing system throughput.

Non-collaborative Allocation (NCA). This scheme performs channel allocation at the transmission set level. A collaborative relay set consists of both direct and relay links. For AF-RAKE, each set is assigned with one channel. If there is no channel available, both links are disabled. For AF-FDD, each set is assigned with two channels. If there is no channel available, both links are disabled. If there is only one channel available, only the direct link will be set up. This scheme attempts to maximize system throughput at the cost of blocking some direct links.

In addition to use conflict-free channel allocation, we can also implement CA and NCA strategies using channel contention protocols. In multi-channel MAC protocols such as MMAC, links contend for channels. CA can assign higher priority in terms of shorter contention window, while NCA requires both direct and relay links contend as a set.

V. ASYMPTOTIC ANALYSIS OF COLLABORATIVE RELAY

In this section, we analyze the effectiveness of collaborative relays in large-scale wireless networks. We explore the asymptotic trend assuming a large-scale random wireless network where each node communicates with one of its farthest neighbors at the edge of communication range. We reduce the network into a unit disk graph, and model the penalty of expanded interference by an increase in channel resource usage, and translate it into a penalty of single-link capacity. This penalty can be used to design cooperative relay strategy that takes into account the impact of interference.

A. Unit Disk Graph Model

We start from a brief discussion on the unit disk graph model used for our analysis. Assuming a network of nrandomly deployed nodes, we can build a unit disk graph as follows: each node maps to a vertex, two vertices connect if and only if they are within a distance of d. An alternative way to build this graph is to associate each node with a disk which is centered on the node and with radius d/2. Two vertices connect if and only if the two corresponding disks overlap. A simple geometric argument can justify the equivalency between these two constructions. Additional information on unit disk graph can be found in [28]. Next, we show that one can approximate the interference condition of a collaborative relay network by a unit disk graph.

B. A Simplified Model of Large-scale Random Networks

In Section IV, we show that each collaborative relay or direct transmission has an interference sensitive region. However, the size of the disk depends on the distance between source, relay and destination nodes. To make the analysis tractable, we consider the worst case scenario: each transmission pair is of the maximum distance apart. This produces the largest interference sensitive region. Again, we assume that each node has the same interference-sensitive range r_I .

Under these assumptions, we show that one can model the direct transmission and the collaborative relay transmission using a unit disk graph. To do so, we first choose the center of the link as the disk center. We approximate the interference sensitive region with a large disk of radius X, and the final unit disk has a radius of X/2, which follows from the second construction of unit disk graph described in Section V-A. Next we illustrate the process for both direct transmission and collaborative relay transmission.

In a direct transmission, source node S and destination node D are of distance r_C apart, where r_C is the communication range. For any two links, if their median are $r_I + r_c$ apart, then they will not conflict. Therefore, we can approximate the interference-sensitive region of link S-D by an open disk with radius d_d .

$$d_d(r_C) = r_I + r_C = (\lambda + 1)r_C.$$
 (9)

where $\lambda = r_I/r_C$, hereby referred to as *interference sensitive ratio*. Thus, the disk associated with S-D link has radius $d_d/2$.

In the corresponding conflict graph, each link (S, D) is represented by a disk of radius $d_d/2$ centered at the median location between S and D. d_d represents the maximum separation between the centers of two disks where each transmission will not be disturbed by any transmitter of the other disk. Any two links could conflict if their corresponding disks overlap, *i.e.* the centers of both disks are no more than d_d distance apart.



Fig. 5. Disk Model of Direct Transmission. (a) Interference-sensitive region of direct transmission. (b) Approximation of the region by a large disk. (c) Disk associated with a direct transmission in disk graph model.

Similarly, when S and D are assisted by a relay node R where $d(S,R) = r_1$, $d(R,D) = r_2$, we can approximate the interference-sensitive region as a disk of radius

$$a_{coop}(r_{C}, r_{1}, r_{2}) = \begin{cases} r_{I} + r_{C} & : r_{1}^{2} + r_{2}^{2} \leq r_{C}^{2} \\ r_{I} + \frac{1}{2}r_{C} \left(\sqrt{1 + \frac{(r_{1}^{2} + r_{2}^{2} - r_{C}^{2})^{2}}{4r_{1}^{2}r_{C}^{2} - (r_{C}^{2} + r_{1}^{2} - r_{2}^{2})^{2}}} + 1 \right) : otherwise \end{cases}$$

$$(10)$$

Apparently, the interference-sensitive region depends on choice of r_1 and r_2 . We focus on the worst-case interference scenario where the relay node is of distance ρr_C from both source and destination, where ρ is referred to as *relay distance factor*.² Let $r_1 = r_2 = \rho r_C$ and we can approximate the interference-sensitive region of the relay link with an open disk of *radius* d_{coop} ,

$$d_{coop} = \begin{cases} (\lambda+1)r_C & : \ 1/2 < \rho < \sqrt{2}/2 \\ \left(\lambda + \frac{\rho^2}{2\sqrt{\rho^2 - 1/4}} + \frac{1}{2}\right)r_C & : \ \sqrt{2}/2 \le \rho \le \lambda \end{cases}$$
(11)

Thus, similarly as before, the disk associated with the cooperative relay link is centered at the median of S,D, and R and with radius $d_{coop}/2$.



Fig. 6. Disk Model of Collaborative Relay. (a) Interference-sensitive region of a cooperative relay. (b) Approximation of the region by a large disk. (c) Disk associated with a cooperative relay in disk graph model.

Using the above disk model, we can characterize a large random network using a random disk graph. Each vertex represents a randomly deployed link (direct or collaborative relay), located at the median. Each vertex is associated with a disk of radius $d_d/2$ or $d_{coop}/2$ where if two disks overlap, the corresponding vertices conflict and an edge exists between the two. When a network contains only direct links or only collaborative relay links, the corresponding random disk graph becomes a random unit disk graph.

C. Asymptotic Resource Usage of a Random Network

Using existing results of random unit disk graph [29], we can derive the asymptotic number of channels required to achieve conflict-free transmission.

Theorem 1 (C. McDiarmid [29]): Suppose n disks with diameter d(n) are uniformly distributed in a unit area. Let d = d(n) satisfy $d \to 0$ and $d^2n/\ln(n) \to \infty$ as $n \to \infty$. Let $\chi(n)$ represent the minimum number of channels required to assign each disk with a channel without any conflict. Let

$$k(n) = \frac{\pi}{4}d^2n\tag{12}$$

Then as $n \to \infty$, $\chi(n)/k(n) \to 2\sqrt{3}/\pi$ almost surely.³

We define $\chi_D(n_L)$ as the minimum number of channels required when there are n_L direct transmission links only, and $\chi_C(n_L)$ as the minimum number of channels for n_L collaborative relay links only. Because $\chi(n)$ is a random variable, so are $\chi_D(n_L)$ and $\chi_C(n_L)$. Hence, in the following discussions, arithmetic operations on these variables are all in the context of "almost surely". Using Theorem 1 and AF-RAKE, we have

$$\frac{\chi_C(n_L)}{\chi_D(n_L)} = \frac{r_{coop}^2}{r_d^2} = \begin{cases} 1 & :1/2 < \rho < \sqrt{2}/2\\ \frac{\left(\lambda + \frac{\rho^2}{2\sqrt{\rho^2 - 1/4}} + \frac{1}{2}\right)^2}{(\lambda + 1)^2} & :\sqrt{2}/2 \le \rho \le \lambda \end{cases}$$
(13)

One observation from (13) is that when the relay distance factor ρ is less than $\sqrt{2}/2$, collaborative relay does not result in extra interference. However, the tradeoff is that restricting the relay selection to such small area will limit the possibility of finding a relay. Finally, we recognize that the model simplification in Figure 5 and 6 can heavily impact this conclusion, at least the choice of $\sqrt{2}/2$. Nevertheless, this still represents the advantage of using a nearby relay node. This result can be further refined by using physical interference model.

D. Effectiveness of Cooperative Relay

Under a fixed amount of resource, using extra number of channels implies reduced per-channel resource, which we refer to as the penalty of increased interference. In the following, we integrate the penalty with the collaboration gain and examine the overall system throughput gain for using each collaborative relay strategy over direct transmissions.

Let the system throughput S be the sum of individual user throughput. Because each transmission is of identical length, the overall system throughput is the product of throughput per-resource and the number of transmissions.

$$S_{DT}(n_L) = n_L \cdot C_{DT} \cdot \frac{B}{\chi_D(n_L)}$$

$$S_{AF-RAKE}(n_L) = n_L \cdot C_{AF-RAKE} \cdot \frac{B}{\chi_C(n_L)}$$

$$S_{AF-FDD}(n_L) = n_L \cdot C_{AF-RAKE} \cdot \frac{B}{2\chi_C(n_L)} (14)$$

where *B* represents the total amount of available resource, and C_{DT} , $C_{AF-RAKE}$, C_{AF-FDD} represent the capacity of direct transmission, AF-RAKE and AF-FDD defined in Section II. As mentioned before, AF-FDD requires 2× channels compared to AF-RAKE.

For AF-RAKE based collaborative relay to be effective, *i.e.* having a higher system throughput compared to that of direct transmission, we require

$$\frac{S_{AF-RAKE}(n_L)}{S_{DT}(n_L)} \ge 1.$$
(15)

Equivalently, the *collaboration gain*, defined as $\frac{C_{AF-RAKE}}{C_{DT}}$, needs to overcome the interference degradation, *i.e*

$$\frac{C_{AF-RAKE}}{C_{DT}} \ge \frac{\chi_C(n_L)}{\chi_D(n_L)} := T_{AF-RAKE}.$$
 (16)

²In other words, in real system a node chosen to be relay node must be less than ρr_C apart from its source and destination nodes. Generally, the relay distance factor ρ should be no great than λ . On the other hand, it should not be too small, otherwise, the link will only have very small probability to find a relay and thus the system performance will be degraded.

³"Almost surely" is a probability theory terminology, which means that as n goes to infinity the event happens with probability 1 [30, Ch.6].

We hereby refer to $T_{AF-RAKE}$ as the *critical ratio* of AF-RAKE. Similarly for AF-FDD, we require

$$\frac{C_{AF-FDD}}{C_{DT}} \ge \frac{2\chi_C(n_L)}{\chi_D(n_L)} := T_{AF-FDD}.$$
 (17)

Substitute (13) into (16) and (17), we have

$$T_{AF-RAKE} = T_{AF-TDD} = T_{DF}$$

$$= \begin{cases} 1 & : 1/2 < \rho < \sqrt{2}/2 \\ \frac{\left(\lambda + \frac{\rho^2}{2\sqrt{\rho^2 - 1/4}} + \frac{1}{2}\right)^2}{(\lambda + 1)^2} & : \sqrt{2}/2 \le \rho \le \lambda \end{cases}$$
(18)

$$T_{AF-FDD} = \begin{cases} 2 : 1/2 < \rho < \sqrt{2}/2 \\ \frac{2\left(\lambda + \frac{\rho^2}{2\sqrt{\rho^2 - 1/4}} + \frac{1}{2}\right)^2}{(\lambda + 1)^2} : \sqrt{2}/2 \le \rho \le \lambda \end{cases}$$
(19)

Table II lists some numerical values for the critical ratio. For example, when $\rho = \lambda = 1$, $T_{AF-RAKE} = 1.0788$ and $T_{AF-FDD} = 2.1577$. This conclusion is different from those from single link transmission where collaboration always benefits (see Section II). The critical ratio provides a guideline for collaborative relay design on a single link basis.

E. Extensions to Arbitrary Number of Relays

The above conclusions assume each direct link is assisted by exactly one relay. When there are multiple relay nodes, the impact of interference depends on the node that is furthermost from the source and destination. Under the constraint that each relay node needs to be within $r_R = \rho r_C$ from source and destination, the maximum interference generated by the collaborative relay is represented by a disk of radius

$$d_{max} = \begin{cases} (\lambda + 1)r_C & : \quad 1/2 < \rho < \sqrt{2}/2 \\ \left(\lambda + \sqrt{\rho^2 - 1/4} + \frac{1}{2}\right)r_C & : \quad \sqrt{2}/2 \le \rho \le \lambda \end{cases}$$
(20)

Hence, the corresponding disk in unit disk graph is of radius $d_{max}/2$.

Such simplification allows us to approximate the *maximum* critical ratio T_{max} as following, which represents the critical ratio when the multiple cooperative relays are simultaneously chosen and placed in positions that lead to the maximum interference sensitive range.

$$T_{max} := \frac{\chi_{MAX}(n_L)}{\chi_D(n_L)} = \begin{cases} 1 & : 1/2 < \rho < \sqrt{2}/2\\ \frac{(\lambda + \sqrt{\rho^2 - 1/4} + 1/2)^2}{(\lambda + 1)^2} & : \sqrt{2}/2 \le \rho \le \lambda \end{cases}$$
(21)

Intuitively, a direct link should use collaborative relay if the relay throughput gain exceeds T_{max} . For $\rho = \lambda = 1$ $T_{max} = 1.3995$ if using AF-RAKE or AF-TDD, and $T_{max} = 2.7990$ if using AF-FDD or DF. Table II lists additional set of results under various choices of ρ and $\lambda = 1$. However, this simplified analysis is based on the assumption that all the nodes still have the same interference-sensitive range. When there are multiple relays, the transmit power density changes and thus the interference model is required to determine the relationship between r_I and the number of relay nodes, which we will address in a future study.

VI. SIMULATION RESULTS

A. Simulation Setup and Evaluation Metrics

Our simulations are conducted under the assumption of noisy immobile radio networks, where each transmission experiences frequency flat Rayleigh fading. We focus on random networks by randomly deploying a set of nodes on a given area and examine the average results over 1000 deployments⁴

We use a simple minimum-distance-idle-first strategy to select relay node, where each source node selects a nearest (to both source and destination) idle node as its relay. We focus on conflict-free transmissions and apply the algorithms from [25] to allocate channels to links but vary the priorities of direct and relay links. We consider several performance metrics.

- *System throughput.* We examine total system throughput based on both ergodic and outage capacities.
- System Performance Gain. It is an indicator on how much a collaborative relay should achieve (without any impact of interference) in order to remain advantages in an interference-oriented environment.
- *Link blocking probability.* It represents the probability of direct link failures due to channel shortage. A direct link fails if it is not assigned with any channel.
- *Relay usage probability.* It measures the probability of relay usage across all the *active* direct links.

While system throughput and performance gain measure the overall system performance and the superiority of collaborative relay, link blocking probability measures the impact on individual user fairness.

We conduct simulations in two different scenarios. In the first scenario, we assume that network resource can be flexibly partitioned and per-channel resource depends on the total number of channels required for conflict-free transmissions. This allows us to verify the analytical findings developed in Section V. In the second scenario, we assume that network resource is pre-partitioned into fixed number of channels, and apply the channel allocation schemes proposed in Section IV to examine the tradeoffs between throughput and fairness.

B. Scenario 1: Flexible Channel Partitions

We start with random networks with a fixed amount of spectrum but have freedom in channel partition. The goal of access is to partition spectrum so that each link is assigned with exact one channel. We randomly deploy nodes in a unit disk area where each node has a communication range of $r_C = 0.1$. In addition, we assume each link experiences Rayleigh flat fading with $\gamma = 0dB$ and $E(|h_{ij}|^2) = 1$. For link pairs within smaller distance, we scale their channel statistics following the propagation model defined in [21], with the pathloss factor $\alpha = 2$. For each direct link, we randomly deploy one node that is within $r_R = \rho r_C$ distance from both source and destination as the relay node, to capture the randomness of neighboring nodes' availability.

⁴We also examined the performance under grid networks, with similar conclusions. The results are omitted due to space limitation.

	TABLE II	
NUMERICAL	VALUES OF CRITICAL	Ratio $(\lambda = 1)$

	ρ	$\leq \sqrt{2}/2$	0.75	0.80	0.85	0.90	0.95	1
One Node	AF-RAKE	1	1.0031	1.0124	1.0257	1.0416	1.0595	1.0788
Relay	AF-FDD	2	2.0062	2.0249	2.0514	2.0833	2.1190	2.1577
Multiple Nodes	AF-RAKE	1	1.0599	1.1284	1.1962	1.2637	1.3315	1.3995
Relay	AF-FDD	2	2.1198	2.2567	2.3923	2.5275	2.6629	2.7990

TABLE III The Number of Channels Required

No. of L	nks	50	100	150	200	300	400	500
DT-F	EXP	3.24	4.42	5.38	6.22	7.77	9.27	10.64
DT-A	NA	1.73	3.46	5.20	6.93	10.40	13.86	17.32
AF-I	EXP	3.44	4.81	5.90	6.87	8.76	10.53	12.18
AF-A	NA	1.87	3.74	5.61	7.47	11.21	14.95	18.69

We first verify the analytical results in Section V by comparing the number of channels required obtained through simulations to those from asymptotic analysis. Table III summarizes the results for both direct transmission (DT) and AF-RAKE, where EXP represents simulation results and ANA represents analytical results. In this example, we assume $\lambda = \rho = 1$. We observe that the analytical results provide a reasonable approximation to more realistic simulation results. These results clearly demonstrate the impact of expanded interference on collaborative relay.

Next, we examine the gain of collaborative relay in terms of system throughputs. In Figure 7 we examine the system performance gain $S_{AF-RAKE}(n_L)/S_{DT}(n_L)$ in terms of both ergodic and outage throughputs, assuming $\lambda = \rho = 1$. When the number of links is small, *i.e.* less than 100, the impact of interference is negligible and the system performance gain is similar to single link throughput gain, i.e. 1.2 for ergodic throughput, and 1.8 for outage throughput. As the number of active links increases, the system performance gain decreases as the impact of interference starts to dominate. This demonstrates the relation between collaboration gain and interference degradation. We also include the asymptotic gain based on the worst case scenario where each link pair is of r_C apart, an upper-bound to system performance gain under large number of links. Note that the upper-bound is only valid for large number of links, since it is asymptotic.

In Figure 8, we examine the impact of relay distance factor ρ assuming 300 links and $r_C = 0.1$. It is not surprising that increasing ρ leads to system degradations. However, the value of ρ^2 is directly proportional to the probability of finding an available relay node, which is not captured in this simulation.

C. Scenario 2: Fixed Channel Partitions

Next, we examine the performance of the proposed channel allocation schemes in Section IV on random networks. We deploy 1000 nodes in a unit disk area and randomly set up a number of links. This simulation is different from those in previous section - there is no guarantee that each link can find an available relay. We divide the spectrum into a fixed number of channels, each of bandwidth W = 1. We allocate channels to both direct and relay links to avoid conflict and attempt to



(b) Outage Throughput

Fig. 7. The effectiveness of collaborative relay in terms of system performance gain, assuming random networks, $\rho = \lambda = 1$.

maximize the number of links supported. Similarly, each node has a communication range of $r_C = 0.1$.

Figure 9 shows the average system ergodic and outage throughputs, for direct transmissions (DT), AF-RAKE and AF-FDD. We also compare the performance of collaborative and non-collaborative channel access, denoted as CA and NCA. In this example, we assume the system has two orthogonal channels with independently identical channel statistics. We observe that AF-RAKE significantly outperforms AF-FDD and DT in terms of both ergodic and outage throughputs. Similarly to the observation in Figure 7, the overall system



Fig. 8. The effectiveness of collaborative relay in terms of system performance gain, assuming random networks, 300 links.

performance gain decreases with the number of links. For outage throughput, the gain decreases from roughly 200% at 10 links to 160% at 300 links. AF-FDD outperforms DT in terms of outage throughput but has very marginal improvement in ergodic throughput using CA channel access. This is due to the fact that AF-FDD requires two channels for each collaborative relay set while AF-RAKE requires one.

One interesting observation is that NCA outperforms CA in outage throughput but not in ergodic throughput. This can be explained by the statistics in Figure 10. By prioritizing direct link over relay link, CA leads to smaller probability of direct link failure, but at the cost of less relay usage. In the case of outage throughput, the loss due to reduced relay usage exceeds the gain of supporting more direct links. Therefore, NCA based channel access outperforms CA. However, the choice between CA and NCA depends on target applications. CA focuses on minimizing link blocking rate and NCA focuses on system throughput.

Next, we examine the impact of the amount of resource on system performance. We assume that each channel has bandwidth W = 1 so that adding channels implies more spectrum. Figure 11 illustrates the system throughputs as-



Fig. 9. System ergodic throughput and outage throughput as a function of

suming there are 160 active direct links in the network. As expected, increasing the number of channels significantly improves system throughputs, especially the system outage one. These observations are explained by the statistics of direct link failure and relay usage in Figure 12.

VII. DISCUSSIONS

A. Extension to mobile scenarios

the number of links for two channels.

In this paper, we focus on static networks such as sensor or mesh networks. However, our work can be extended to mobile networks. In particular, because our analytical results assume random networks with a fixed network density, the results can be easily applied to a large-scale mobile network where nodes move but maintain similar network density.

We also note that there are several existing solutions on distributed channel allocations where nodes apply local actions to tune network-wide channel allocations [26], [27]. Our approach can be extended to include these new advances. However, how to reliably identify conflicting neighbors remains as a challenging open research problem.



(b) Probability of relay usage

Fig. 10. Cooperation usage and link success statistics for systems with two channels

B. Extension to other Relay Strategies

In this paper we examine the performance of collaborative relay by focusing on amplify-and-forward (AF) strategies. Our approach can be extended to other strategies by deriving the appropriate interference-sensitive range, and throughput without considering interference. For example, multi-source-multirelay-multi-destination scheme in [11] needs more channels or time slots compared with AF. As a result, the number of channels required for conflict-free transmissions will increase compared to that of AF, which leads to a higher critical ratio in terms of single-link throughput improvement. Overall, the analytical result provides a practical guideline to design and evaluate cooperative relay without delving into complex network optimization.

C. Other Practical Issues

In this paper, we use a simplified protocol implementation. In practice, the performance of collaborative relay also depends on the coordination overhead, level of synchronization, node mobility and channel access protocols. We are currently



Fig. 11. System ergodic throughput and outage throughput as a function of the number of channels assuming 160 active direct links.

researching on the impact of these issues by implementing collaborative relay on 802.11 devices.

VIII. CONCLUSION

In this paper, we examine the feasibility of applying collaborative relay to large-scale wireless networks for throughput improvement. While collaborative relay is proven to be effective in single link transmissions, its effectiveness in largescale networks is seriously penalized by the increase in interference. We show that asymptotically there exists a threshold critical ratio, where the collaborative relay improves system throughput if the corresponding collaboration gain in single link transmission exceeds that critical ratio. Simulation results verify the accuracy of asymptotic estimations. It is clear that careless usage of cooperative relays in an interference-limited environment could lead to higher connection blocking probability. However, opportunistic relay usage when neighboring nodes are idle can lead to higher link throughput. The system should apply a fair and efficient resource allocation scheme to balance such interdependency.

While our work provides an initial step towards examining collaborative relay in a large network, there are still many



(b) Probability of relay usage

Fig. 12. Cooperation usage and link blocking statistics for variable numbers of channels and 160 links.

open problems. In this paper we only examine the performance of a limited set of collaborative relay strategies, focusing on amplify-and-forward (AF). However, the emphasis is on the number of channels that a collaborative relay uses. We plan to explore other strategies with additional patterns of channel usage in future work. In addition, we use a simplified system implementation without considering the impact of protocol overhead, synchronization, node mobility and channel access protocols. We are currently researching on the impact of these issues by implementing collaborative relay on 802.11 devices. Finally, our analytical results are derived based on the protocol interference model which is a pessimistic simplified version of the physical interference model. Our future work is to extend this study based on the physical interference model and examine the implication of power allocation/control on system performance.

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