

A Novel Link Scheduling Algorithm for Wireless Networks using Directional Antenna

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Abstract—For a given set of communication links whose senders transmit at a fixed power level, it is a hot problem to select a maximum set of links that can be transmitted simultaneously, which is known to be NP-hard. The existing algorithm only apply to the condition of omnidirectional transmission. This paper addresses the problem in a plane wireless network where the nodes use directional antennas under physical interference model. We develop a directional interference model applicable to such networks, and first propose an approximation algorithm to solve scheduling problem under directional environment. Besides, we extend this problem to the situation of full-duplex transmission, and give a solution. At last, mathematical analysis and simulations proved the correctness of the algorithm and the great advantages of directional antenna.

Index Terms—Link Scheduling, Directional Antennas, Physical Interference Model

I. INTRODUCTION

As we all know, link scheduling is the most fundamental and typical problem which can significantly influence network performance. Since Gupta and Kumar first studied on the methods to schedule links and enhance the capacity in Wireless Sensor Networks [1], lot of problems were proposed one after another. In the case of shared only one channel, a plurality of communication links transmission at the same time will lead to signal interference. In wireless network, link scheduling problem is to reduce the maximum interference when the links transmit simultaneously, and ensure the message can be successfully accepted by the receiving node. The main objective is to achieve efficient spatial reuse and increase network capacity, considering wireless interference among concurrently transmitting nodes.

The problem maximum link scheduling (MLS) is to seek a largest set of links from a given set A that can be scheduled simultaneously. Given a set of communication link requests $L = \{l_1, l_2, \dots, l_n\}$, with l_i denoting the i th link request. MLS intends to find a maximum subset of links $S \in L$ to be scheduled simultaneously, designated to one time slot, given a set of communication links with each having a unit traffic demand. This optimization problem is NP-hard proved by Goussevskaya [2]. There are a lot of studies to this aspect, some better performance approximation scheduling algorithms have been proposed [3], [4].

By concentrating the energy in specific direction, opposed to omni-directional transmission, directional antennas can provide the benefits of increased range, reduced interference and

increased spatial reuse of bandwidth. We can benefit from the advantages provided by directional antennas if the scheduling does not take into account the nature of the beam formation at each node.

In order to solve this problem, there are mainly two kinds of interference model are used. One of them is graph-based model [5], including the protocol interference model, CTS/RTS model, K-hop interference model and so on. By localizing the interference of a transceiver on others, it is much easier to track and handle. Unfortunately, the graph-based interference model is too idealistic and simple so that there are many vital factors are ignored. Another kind of interference model is physical interference model [1], which is also known as SINR (signal-to-interference-plus-noise-ratio) model. In physical interference model, the global interference is considered and is more related to physical layer constraints. It is a challenging issue to design algorithms under the SINR model. Most of the researches are based on graph-based models which is too optimistic. To make our research more accurate, we adopted physical interference model in this paper.

As time goes on, much more conditions that will influence link scheduling performance significantly have been taken into account in the research, like power control which is a vital element. Many kinds of power control for instance, linear transmission power control, uniform transmission power control and arbitrary transmission power control are considered [6], [7]. Another important issue is about distribution or centralization problem. Most studies focus on the centralized implementation, but some develop distributed even localized link scheduling algorithms with provable throughput performance [8].

In this paper, we apply directional antennas to the research of MLS problem based on the physical interference model, which can significantly reduce ambient interference and ensure more nodes can be transmitted at the same time. Then we also take full-duplex transmission into consideration, in which every node is both sender and receiver. It is more complex but also more realistic. The main contributions can be presented as follows.

- We consider a more realistic and accurate SINR interference model, and design a directional interference model applicable to directional environment. In this model, we take directional angle and antennas gain into account, and provide a more realistic interference model which

is suited for directional antennas.

- We study link scheduling to maximize the network throughput under directional environment. We propose an approximation algorithm with uniform power assignment. Our performance analysis and simulation study demonstrate the efficiency of LSDA (Link Scheduling Algorithm based on Directional Antenna) and FLSDA (Full-duplex Link Scheduling Algorithm based on Directional Antenna) compared to the state-of-the-art algorithms.
- This paper we consider the situation of both half-duplex and full-duplex, and gives corresponding algorithm for each situation. The design of algorithm was motivated by cellular networks, which is to find maximum optimal cells and avoid interference from the neighboring cells.

The rest of the paper is organized as followed. We describe related work in Section II. Section III give our directional interference model. In section IV, we present scheduling algorithm for the problem of MLS and Full-duplex MLS, also provide mathematical analysis for the algorithm. Section V show the simulation results to illustrate the performance of our scheduling algorithm, and section VI concludes the paper.

II. RELATED WORK

Since Gupta and Kumar first studied on the methods to schedule links and enhance the capacity in Wireless Sensor Networks, three sub-problems were proposed one after another: maximum link scheduling (MLS) problem [9], maximum weighted link scheduling (MWLS) problem [10] and shortest link scheduling (SLS) problem [11]. In the process of studying the scheduling problem, some other factors are also need to take account of, such as the interference, power control, centralized or distributed scheduling, omnidirectional or directional antennas and so on. Sometimes we also should take into account both links scheduling and other problem, links channel assignment or rate adaptation [12].

Goussevskaia et al. [2] give us a simple proof that link scheduling under the SINR model is NP-hard. He also proposed an $O(l_{max}/l_{min})$ factor approximation algorithm for MLS problem with a uniform power assignment, where l_{max} and l_{min} denote the length of the longest and the shortest link, respectively. The algorithm used greedy strategy to construct a scheduling in different classes partitioned by length. After a while, Goussevskaia [13] made huge efforts on developing a constant approximation bound in the literature and proposed a $O(\log n)$ approximation for the problem of maximizing the number of links scheduled in one time-slot scheduling. Then a factor of $O(1)$ -approximation ratio algorithm was put forward by Halldrsson and Mitra [14]. They extended the transmission power to oblivious power assignment (including uniform, mean, and linear power assignment). Furthermore, the algorithm is applicable for both unidirectional and bidirectional links. In this paper, we will analyze bidirectional communication because bidirectional communication is more consistent with WSNs. Fanghanel et al. introduced the bidirectional version of the scheduling problem and gave a first polynomial time approximation factor algorithm for SLS using

the mean power assignment in general metrics [15]. Now Peng and his team [16] present a new paradigm for SLS problem. This problem was be deep analyse from both theory and application.

The links scheduling algorithms under SINR model are more accurate than that under the graph-based models since the physical interference model can accurately describe the interference in wireless networks. Considering the interference model used in Goussevskaia et al. (2007, 2009)[1,3], it is an approximation of the SINR model, but the effect of noise is ignored in this paper. SINR is simplified to SIR problem after neglecting the ambient noise, in which the transmission scope of a link will become infinite. In this way, the possible number of link classes divided by length is infinite as well. Blough et al. [17] proposed the first SLS algorithm under the exact SINR model. He defined a class of links named black-gray links, whose lengths are equal or near to the maximum transmission scope of the sender. In the extreme case, in which all the links to be scheduled are black-gray, they obtained an upper bound on the length of SLS algorithm, which is $O(|C_0| + \Delta_{max})$, where $|C_0|$ is the number of links in class C_0 , and Δ_{max} is the maximal number of receivers in a cell of class C_k . In [18] and [19], Li proposed a hypergraph model combining graph-based model and SINR models, and demonstrated that the hypergraph model contains the advantages of both graph-based model and the SINR model while avoiding their drawbacks. Namely, the hypergraph model can emulate cumulative interference constraints as hyperedge, where each hyperedge is a set of links that are not allowed to transmit simultaneously. Thus, this model can avoid the defects of graph-based models which overlook the cumulative interference.

The algorithms mentioned above adopt uniform power assignment. Many algorithms for link scheduling with non-uniform power assignment are also addressed by people. Kozal et al. [20] proved the problem of power control and scheduling to be NP-complete by using a reduction from integer programming under the assumption that the values of gain matrix can be chose arbitrarily. In Kesselheim [21], Kesselheim studied the capacity maximization problem with power control in a wireless network. With the objective of maximizing the number of simultaneous links to communicate, he developed an algorithm which selected a subset of links and chosen a power level for each link. He also proved that the algorithm achieves $\Omega(n)$, $\log(n)$ approximation in a fading metric and a general metric space, where n is the number of links.

All the works reviewed by now are centralized, and the studies dont give us a strategy to develop a distributed version. Le et al. [22] proposed a distributed greedy maximal link scheduling algorithm under interference localization. A link l only performs scheduling coordination inside a circle area named interference neighborhood of the link. However, the trivial procedure for determining the interference neighborhood is centralized. Moreover, link l and other links need to calculate their cumulative interference in an iterated procedure, which is impractical in a large scale network. Pei and

Vullikanti [23] proposed a local distributed scheduling and power control algorithm under the SINR model, achieving an $O(\log^2 n)$ approximation factor in the throughput region. Note that the uniform power assignment for all links was adopted, and thus the links in the same link class were assigned the same power. However, they do not give reasons why the number of active nodes in the vicinity of any active node decreases by half after each phase with a high probability.

However, mostly studies of link scheduling based on omnidirectional transmission. Ramamurthi [24] proposed a generalized physical interference model applying to the directional antennas, both taking into account the main lobes and the side lobes of antennas. The benefit of directional antennas to improve network capacity has been deeply analyzed in [25]. Although directional antennas has been studied for cellular networks and has been deployed for cell-sectoring, it is rarely used for OSML problem.

III. THE DIRECTIONAL INTERFERENCE MODEL

An essential issue to link scheduling problem in wireless networks is the interference model. Here is a list of some notation meaning. All the networking nodes V lie in plane and transmit at a fixed power P . The Euclidean distance between any pair of nodes is denoted by d_{uv} . The antenna gain of the node u is G_u . The path loss model is then determined by the path-loss exponent α , which is a constant greater than 2 but less than 6 typically. Specifically, when a node u transmit a signal at power P , the power of this signal is captured by another node v is $Pd_{uv}^{-\alpha}G_uG_v$. The signal quality perceived by receiver is measured by the SINR (signal to interference and noise ratio), which is the quotient between the power of the wanted signal and the total power of unwanted signals and the ambient noise (both internal and external).

Here is a brief introduce to independent set. It is a set of disjoint links that can be scheduled at a same time slot, formulated as follows. A set of links $S = \{(u_1, v_1), (u_2, v_2), \dots, (u_k, v_k)\}$ are said to be independent if:

- (1) all links in S are disjoint;
- (2) for each $1 \leq i \leq k$, the SINR of the link u_i, v_i with respect to $u_j : 1 \leq j \leq k, j \neq i$ is at least β .

1) *physical interference model*: In this model, the transmission from sending node s_v is successfully received by receiver r_v if and only if

$$\text{SINR} = \frac{Pd_{vv}^{-\alpha}}{N + \sum_{l_w \in S} Pd_{wv}^{-\alpha}} \geq \beta, \quad (1)$$

where $\alpha > 2$ is the path-loss exponent, $\beta > 1$ is the minimum SINR required for successful reception and determined by the hardware conditions, N is the ambient noise around receiving node (note that α, β, N are all constants). This models a situation where the SINR must be no less than certain threshold β , in order to correctly interpret the wanted signal. It is a very effective model for omnidirectional transmission, takes the total interference into account due to all transmissions.

In our search, each nodes sending by directional antenna but received by omni-directional antenna. The problem of

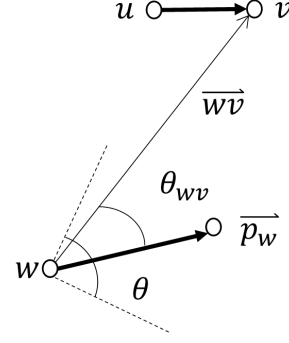


Fig. 1. Angle Judge.

OSML can be formulated as follows. Given a set of links $L = \{l_1, l_2, \dots, l_n\}$, where each link l_i represents a communication request from a sender node s_i to a receiver r_i , and our object is to seek a maximum independent set in one time-slot, that means each links in the independent set can be scheduled simultaneously.

We assume the length of link l_v is denoted by d_{vv} , and simplify directional antennas as only have one main beam, fixing the direction of the antenna. Which the interference range only in a particular angle generated by sending node.

2) *directional interference model* [24]: The link $l_v = (s_v, r_v)$ successfully transmitted under the following conditions:

- (1) In the absence of other links:

$$\text{SNR} = \frac{PG_{s_v}G_{r_v}d_{vv}^{-\alpha}}{N} \geq \beta. \quad (2)$$

- (2) Presence of interference from other links:

$$\text{SINR} = \frac{PG_{s_v}G_{r_v}d_{vv}^{-\alpha}}{N + \sum_{l_w \in S} PG_{s_w}G_{r_v}d_{wv}^{-\alpha}\varphi_{wv}} \geq \beta, \quad (3)$$

where P is transmission power of the transmit node, link $l_w \in S$ scheduled concurrently with l_v , G_{s_v} and G_{r_v} is the antenna gain of sending node and receiving node, it is a constant which deceive by hardware conditions.

The model was proposed to narrow the interference angle by using directional antenna, so we need to determine whether exist interference between the links that transmission in the same time. Here we use φ_{wv} to solve this problem, where vector \vec{p}_w present the send direction of node w , the send angle of directional antennas is θ_w . Definition as follows:

$$\varphi = \begin{cases} 0, & \theta_{wv} > \theta_w/2, \\ 1, & \theta_{wv} \leq \theta_w/2, \end{cases}$$

where θ_{wv} is the angle between antenna direction \vec{p}_w of sender node w with the vector \vec{wv} that the direction of sender node w to receiver node v . We use this angle to judge whether the receiving node is in the send range of the sender node. As shown in Fig. 1. We can use the following formulate to get

the angle θ_{wv} , and decide the interference between links.

$$\theta_{wv} = \arccos\left(\frac{\vec{p}_w \vec{w} \vec{v}}{|\vec{p}_w| |\vec{w} \vec{v}|}\right).$$

In this research, we assume that all nodes transmit with the same power level P . There are some definition, use $P_{vv} = PG_s G_r / d_{vv}^\alpha$ represent the signal receive power of r_v send by s_v , and $I_{wv} = PG_{s_w} G_{r_v} \varphi_{wv} / d_{wv}^\alpha$ denote the signal interference received by node r_v from a sender s_v that transmit at the same time.

IV. THE SCHEDULING ALGORITHM

In order to solve the scheduling problem of MLS, we present the follow LSDA (Link Scheduling Algorithm based on Directional Antenna) algorithm. Start with some definitions, the *relative interference* (RI) of a link l_u on link l_v , namely $RI_u(v) = I_{uv} / P_{vv}$. The *affectedness* (proposed in [?]) of link l_v , caused by a set of links S , is the sum of the relative interferences of the links in S on l_v , as well as the effect of noise, scale by β , or

$$\begin{aligned} A_S(l_v) &= \beta \left(\frac{N}{P_{vv}} + \sum_{l_u \in S} RI_u(v) \right) \\ &= \beta \frac{\sum_{l_u \in S} I_{uv} + N}{P_{vv}}. \end{aligned} \quad (4)$$

The formula of *affectedness* was got from $\text{SINR} \geq \beta$. Observe that the *affectedness* of link l_v satisfy $A_S(l_v) \leq 1$, equivalent to the $\text{SINR} \geq \beta$, means link l_v can successful transmission.

Our approximation algorithm for LSDA is outlined above. We can simplify the algorithm as a brute-force method. Let L be the set of given communication links, we assume that each link $l_v \in L$ can success communication, added to the solution, its safety ($\text{SINR} \geq \beta$) is guaranteed in the step of select. Set L is a sequence which sorted in the increasing order of length of links. Depend on the definition of SINR, we can get the feature that the shorter the length, the more stable of the link. S stores the link in one of independent set, and I is the set of links have interference with the select link l_v in each iterative. In every iteration of the algorithm, there have three step to select the legal and remove illegal link. The first link l_v in L is moved from L to S , and use this link as the begin of the first round of selection. The first step (line 6) we need discards all links $l \in I$ whose sender are close to the receiver of l_v , meaning $d(r_v, s_w) \leq c \cdot d_{vv}$ (c is a constant bigger than 2, and explained in next part).

$$c = \max \left(2, (2 \cdot 3^3 \cdot \beta \cdot \frac{\alpha - 1}{\alpha - 2})^{\frac{1}{\alpha - 2}} \right). \quad (5)$$

Then make sure the distance between any two links from set I is bigger than d_{vv} , remove the illegal links from L (line 16). The last step (line 23), all links $l_u \in L$, whose *affectedness* $A_s(l_u)$ rose to or above a threshold of $2/3$, are removed from L (the number of $2/3$ will explain in next part). This iterative is repeated until all links in L have been select or deleted. In the end, we will get the scheduling set S from LSDA algorithm.

Algorithm 1 LSDA Algorithm

Require: Set of links in the increasing order of length $L =$

l_1, l_2, \dots, l_n ; $S \leftarrow \emptyset$, $I \leftarrow \emptyset$, $L_1 \leftarrow \emptyset$.

Ensure: OMSL schedule S .

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1: Set  $c$  according to Eq. (5).
2: repeat
3:    $l_v = (s_v, r_v) \leftarrow$  the first link in  $L$ .
4:    $L \leftarrow L \setminus l_v$ ,  $S \leftarrow S \cup l_v$ .
5:    $L_1 \leftarrow l_u \in L : d(s_u, r_v) \leq c \cdot d_{vv}$ ,  $\varphi_{uv} = 1$ .
6:    $L \leftarrow L \setminus L_1$ .
7:   repeat
8:      $l_w \in L \leftarrow$  satisfy  $\varphi_{wv} = 1$ .
9:      $I \leftarrow I \cup l_w$ .
10:  until  $L$  is traversal end.
11:  repeat
12:     $l_a = (s_a, r_a) \leftarrow$  the first link in  $I$ .
13:     $I \leftarrow I \setminus l_a$ .
14:    repeat
15:       $I_a \leftarrow l_u \in I : d_{ua} = d(s_u, s_a) \leq c \cdot d_{aa}/2$ .
16:       $I \leftarrow I \setminus I_a$ .
17:       $L_v \leftarrow I_a \cup L_2$ .
18:    until  $I$  is traversal end.
19:  until  $I = \emptyset$ .
20:   $L \leftarrow L \setminus L_v$ .
21:  Set  $I \leftarrow \emptyset$ .
22:   $L_2 \leftarrow l_u \in L : A_S(l_u) \geq 2/3$ .
23:   $L \leftarrow L \setminus L_2$ .
24: until  $L = \emptyset$ .
25: return  $S$ .
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Next we prove that the obtained schedule is both correct and competitive.

A. Correctness of Link Scheduling Algorithm

In this section we prove that the solution S obtained in LSDA Algorithm is correct, all selected links can be scheduled concurrently without collisions, $\forall l_v \in S$, $A_S(l_v) \leq 1$.

There some definitions need to be used in the proof. For link $l_v \in S$, let S_v^- be the set of links that length shorts than l_v , and S_v^+ be the links longer than l_v , there have some interference between l_v and any link l_w from S_v^- or S_v^+ ($\varphi_{wv} = 1$).

We can get from third part in iteration of the algorithm (line 22), each link in the scheduling set S satisfy that the *affectedness*, $A_{S_v^-}(l_v) \leq 2/3$, means that for each links $l_v \in S$, when the link is add to the set S , the *affectedness* of l_v get by S_v^- is less than $2/3$, since it has not been deleted by in the previous step. In order to ensure that each links in S can be successful communication at the same time slot, the SINR of each link should be satisfy $\text{SINR} \geq \beta$. It show that we just need to ensure $\forall l_v \in S$, $A_{S_v^+}(l_v) \leq 1/3$.

In order to give a clear analyse, here are some geometric definition now. We used D_w present the discs of radius $d_{ww}/2$ around receiver node $s_w \in S_v^+$. From the first elimination criterion, we know the discs D_w do not contain any sender $s_z \neq s_w$ and $s_z \in S_v^+$. Focus on the links set I which have

interference on link l_v . At the first, division the sender set in I into concentric rings $Ring_k$ which have evenly spaced of cd_{vv} around the receiver r_v . Each ring $Ring_k$ contains all senders $s_w \in S_v^+$, for which $k(cd_{vv}) \leq d_{wv} \leq (k+1)(cd_{vv})$. Because of $d_{wv} \geq c \cdot d_{vv}$, so that the first ring $Ring_0$ does not contain any sender from S_v^+ . Consider all senders $s_w \in Ring_k$, for the concentric rings $Ring_k$, $k > 0$. All discs D_w of radius $c \cdot d_{vv}/2$ around node s_w which located in $Ring_k$ must be completely contained in an extended ring $EXRing_k$, and the area is calculated by the fellow formula:

$$\begin{aligned} A(EXRing_k) &= [(d_{vv}(k+1)c + d_{vv} \cdot c/2)^2 \\ &\quad - (d_{vv}kc - d_{vv} \cdot c/2)^2]\pi \\ &= c^2 \cdot 2k \cdot d_{vv}^2 \pi. \end{aligned}$$

Since that each around discs D_w of area $A(D_w) \leq c^2 \cdot d_{vv}^2 \pi/4$ around senders $s_w \in I$ do not intersect, and the minimum distance between r_v and s_w is $k \cdot c \cdot d_{vv}$, $s_w \in Ring_k$, $k > 0$. The total interference coming from ring $Ring_k$, $k > 1$ is bounded by

$$\begin{aligned} I_{Ring_k}(l_v) &\leq \sum_{s_w \in Ring_k} I_{s_w}(l_v) \\ &\leq \frac{A(EXRing_k)}{A(D_w)} \cdot \frac{PG_{s_w} G_{r_v} \varphi_{wv}}{(kcd_{vv})^\alpha} \\ &\leq \frac{4 \cdot 2k}{k^\alpha c^\alpha} \cdot \frac{PG_{s_w} G_{r_v} \varphi_{wv}}{(d_{vv})^\alpha} \\ &\leq \frac{1}{k^{\alpha-1}} \cdot \frac{1}{c^\alpha} \cdot \frac{PG_{s_w} G_{r_v} \varphi_{wv}}{d_{vv}^\alpha} \cdot 2^4. \end{aligned}$$

The value of $PG_{s_w} G_{r_v} \varphi_{wv}$ is fixed in above deduce. $A(EXRing_k)/A(D_w)$ represent the maximum number of links have interference to link l_v in the ring $Ring_k$, and we choose $d(s_w, r_v) = k(cd_{vv})$ ensure the interference to l_v is the maximum. Summing up the interferences over all rings yields

$$\begin{aligned} I_{S_v^+}(l_v) &< \sum_{k=1, \dots, n} I_{Ring_k}(l_v) \\ &\leq \sum_{k=1, \dots, n} \frac{1}{k^{\alpha-1}} \cdot \frac{1}{c^\alpha} \cdot \frac{PG_{s_w} G_{r_v} \varphi_{wv}}{d_{vv}^\alpha} \cdot 2^4 \\ &< \frac{\alpha-1}{\alpha-2} \cdot \frac{1}{c^\alpha} \cdot \frac{PG_{s_w} G_{r_v} \varphi_{wv}}{d_{vv}^\alpha} \cdot 2^4, \end{aligned}$$

where the last inequality holds since $\alpha > 2$. This results in affectedness

$$\begin{aligned} A_{S_v^+}(l_v) &= \beta \cdot \frac{\sum_{l_u \in S_v^+} RI_u(v) + N}{P_{vv}} \\ &= \beta \cdot \frac{I_{S_v^+}(l_v) + N}{P_{vv}} \\ &< \frac{\alpha-1}{\alpha-2} \cdot \frac{2^4}{c^4 \alpha} \cdot \frac{1}{P_{vv}} \cdot \frac{PG_{s_w} G_{r_v} \varphi_{wv}}{d_{vv}^\alpha} \\ &\quad + \frac{N \cdot \beta}{P_{vv}} \end{aligned}$$

In order to simplify the analysis, assume that there have no ambient noise $N = 0$, antennas gain and the sending power

of each nodes were fixed value.

$$\begin{aligned} A_{S_v^+}(l_v) &= \frac{\alpha-1}{\alpha-2} \cdot \frac{2^4}{c^\alpha} \cdot \beta \leq 1/3 \Rightarrow \\ c &= \max \left(2, (2^4 \cdot 3 \cdot \beta \cdot \frac{\alpha-1}{\alpha-2})^{\frac{1}{\alpha}} \right). \end{aligned} \quad (6)$$

We have shown that each $l_v \in S$, satisfy $A_S(l_v) = A_{S_v^-}(l_v) + A_{S_v^+}(l_v) \leq 2/3 + 1/3 = 1$, which means that $\text{SINR} \geq \beta$ for each link in scheduling set S . From above mathematical analysis, the value of c was depend on design of third elimination part in algorithm. The judgement of $A_{S_v^-}(l_v) \leq 2/3$ just personal sense, and we proved the correctly.

B. Algorithm Performance Analysis

To analyze the performance of algorithm, we compared the solution ALG of LSDA algorithm to an optimal solution, say OPT . Considering the complexity of directional transmission, we set some special case, and analyse the performance under those case. At first, give some lemma that satisfy the condition of omnidirectional antennas $\theta = 360^\circ$.

lemma 1: Let X be a feasible solution and let l_v be a link in X . The number of sender nodes in X within distance $k \cdot d_{vv}$, $k \geq 1$ of the receiver r_v is at most k^α .

Proof: The relative interference of each sender $l_u \in X \setminus l_v$, where $d_{uv} \leq k \cdot d_{vv}$

$$RI_u(v) = \frac{I_{uv}}{P_{vv}} = \frac{P/d_{uv}^\alpha}{P/d_{vv}^\alpha} = \left(\frac{d_{vv}}{d_{uv}} \right)^\alpha \geq \frac{1}{k^\alpha},$$

which means the maximum number is k^α , otherwise the $\sum_{l_u \in X} RI_u(v) > 1$. Thus the lemma was been proved.

Then, we set other spacial case that communication angle is $\theta = 180^\circ$ of each nodes, and have the same orientation angle. Along with the direction of the orientation angle, the length of links getting shorter and shorter. In this case, the shortest link have no influence to the transmission of other links, which means the last elimination step (line 23) of algorithm have no effect, but get the interference from the larger length of links. ALG_{180} represent the solution of LSDA algorithm, and OPT_{180} is the optimal solution in the same condition.

lemma 2: In k th iteration of the algorithm, we got one link to the solution ALG_{180} and $|OPT_k| \leq (c + (k-1)/2)^\alpha$, OPT_k was the optimal solution in k th iteration.

Proof: Due to the special condition, we just need analyse the first and second elimination part in the algorithm. Consider the set $X_v \subseteq OPT_{180}$ eliminated in the first part of algorithm (line 6), in the iteration when link $l_v \in ALG_{180}$ was added to the scheduling solution. Each link $l_w \in X_v$ is of length at least d_{vv} , and the distance of its sender at most $c \cdot d_{vv}$ from receiver r_v . By Lemma 2, there can be at most c^α in the set X_v .

For the second part of the proof, which equivalent to a pretreatment for the next iteration of the algorithm. Consider the set $L_v \subseteq L$ that the length of each links bigger than the link $l_v \in ALG_{180}$. In this part of eliminated, ensure the distance of each sender of links at least $d_{vv}/2$. We used the Y_k

represent the links which exist in second part of elimination at k th iteration, and the d_k is the distance of the k th iteration when link $l_k \in ALG_{180}$ added to the scheduling solution, $d_k = d(l_k)/2$.

Combine the result of the last iteration. In the k th iteration, it is possible that the links in the discs around $l_w \in Y_k$ was been delete with the radius at most R_k . In the first iteration, the radius $R_1 = d_1 = d(l_1)/2 \leq d_k$. If $k = 2$, $R_2 = R_1 + d_2 \leq 2d_k$. From the induction result, in the k th iteration, which l_v added to the solution. At worst, maybe delete all the links in the discs of radius $R_k = R_{k-1} + d_{vv}/2 \leq k \cdot d_{vv}/2$ around each sender $s_w \in Y_k$.

In general, during the process of k th iteration, $l_v \in ALG_{180}$ was the legal link. Combine the first part and second part of elimination. Each link $l_w \in X_v$ is of length at least d_{vv} and has its sender of distance at most $c \cdot d_{vv} + (k-1) \cdot d_{vv}/2$ from receiver r_v . Therefore, can be at most $(c + (k-1)/2)^\alpha$ senders in X_v .

Due to the complexity of the directional antennas, we can not offered generally demonstration for the algorithm. But the number of non-intersect links is the least number of scheduling links. In other words, all of the non-intersect links can be scheduling at the same time.

C. Full-duplex Algorithm

Nowadays most of the researches about MLS problem are under unidirectional communication model which is more idealized and easy to tackle. However when it comes to the full-duplex transmissions, i.e. each node is able to send and receive signal at the same time, things are more complex. Because all links can behave as the sender and the receiver, the former definition of φ is inappropriate. So we need to redefine some symbols and put forward a new algorithm which is suitable for the full-duplex transmissions situation.

First of all, we redefine the means of φ as:

$$\varphi = \begin{cases} \infty, \theta_{wv} > \theta_w/2, \\ 1, \theta_{wv} \leq \theta_w/2, \end{cases}$$

Then define the distance between the sending node s_v and receiving node r_u as $D_{s_v, r_u} = d(s_v, r_u) \cdot \varphi$, and we redefine the distance between two links l_v and l_u as $D(l_v, l_u) = \min\{D_{s_v, r_u}, D_{s_v, s_u}, D_{r_v, s_u}, D_{r_v, r_u}\}$. The *affectedness* is calculated by the follow formula:

$$\begin{aligned} A_S(l_v) &= \beta \left(\frac{N}{P_{vv}} + \sum_{l_u \in S} RI_u(v) \right) \\ &= \beta \frac{\sum_{l_u \in S} I_{uv} + N}{P_{vv}} \\ &= \beta \frac{\sum_{l_u \in S} PG_{s_u} G_{r_v} D_{l_u, l_v}^{-1} + N}{P_{vv}}. \end{aligned} \quad (7)$$

Before algorithm description, we consider the condition of full-duplex transmissions. If only two links l_v and l_u transmit concurrently, the sufficient condition of link l_u communicating successfully is that $SINR_{min}(l_v) \geq \beta$. Then we give the

definition of "border distance", which is the minimum distance of concurrently transmitting links.

Definition: border distance(BD): a radius depend on the distance of link l_v , which is a circle does not include any others.

$$BD(l_v) = (2^4 \cdot \pi \cdot \beta \cdot \frac{\alpha - 1}{\alpha - 2})^{\frac{1}{\alpha}} \quad (8)$$

By the definition of "border distance", in order to ensure links l_v and l_u can transmit at the same time, the distance between l_v and l_u should satisfy: $D(l_v, l_u) \geq BD(l_v)$.

Next, we will present the algorithm of FLSDA (Full-duplex Link Scheduling Algorithm based on Directional Antenna), which solved the extend problem of MLS. The FLSDA works as follows:

1) First, the input is a set of candidate links $L = \{l_1, l_2, \dots, l_n\}$, which are ranked from shorted to the longest by the distance of link. We choose the first one l_v in the candidate set delete from L , and put it into the output set S as the first solution.

2) Then, in order to make sure $SINR(l_v) \geq \beta$. The selection mechanism delete the link l_u whose distance to l_v is shorter than $BD(l_v)$ by border distance, $D(l_u, l_v) \leq BD(l_v)$.

3) Pick up the links $l_u \in L$ which get interference by l_v as links set I . We judge whether link l_v influence link l_u by formula: $D(l_v, l_u) < \infty$.

4) Choose the shortest link $l_u \in I$ and delete the links whose distance to l_u is shorter than $BD(l_v)$ from link set L and I . Repeat this remove process until the distance between every two links in I is larger than $BD(l_v)$.

5) Next, we decide the inference to links $l_u \in L$ by the current solution. Delete all the links l_u in L which satisfy $A_S(l_u) \geq 2/\beta$.

6) Repeat all above step until the set L is empty, and output the solution which is a set can transmit at the same time.

The process of full-duplex link scheduling algorithm is similar to LSDA. So the correctness has been proved in before. Next section we show the performance of LSDA and FLSDA.

V. EVALUATIONS

In this section, we evaluate the performance of our link scheduling algorithms through simulation analysis, and analyse the influence by the element of links number, directional angle, antennas gains and so on. We also make compared with One-slot scheduling algorithm (OSSA proposed in [13]) to show the performance.

To simplify the analysis of experiment result. We list the following conditions: (1) All of the links random distributed on a plane field of size 1000*1000 units (see Fig. 2). (2) All of the links transmit in the same power. (3) In the experimental, we ignore the influence by ambient noise $N = 0$.

In our link scheduling algorithm, The size of the independent set present the performance of the algorithm. In our experiment, we using the size of output set as the performance standard, and the parameter values used in experiment is present as following (as shown in Table I).

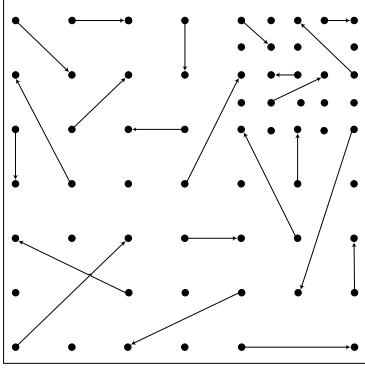


Fig. 2. Simulated Topologies.

TABLE I
DEFAULT CONFIGURATION PARAMETER

Parameter	Description
Field	1000 units
Antenna Gain G	20dBi
Directional Angle θ	120°
Minimum SINR Required β	1.2
Path-loss Exponent α	3
Power P	10w

A. Experimental Results and Analysis

In this subsection, we demonstrate that our algorithms LSDA and FLSDA via simulation study.

Assume that all nodes are deployed in a large network region. The length of links l varies from 10 to 20 unit. Some important parameters are listed before. In our design, we fixed those parameters and increase the number of candidate links set to illustrate the performance of our algorithm. The results are shown in Fig. 3, we analyse the influence by the size of the input set of links, and increase from 100×2^0 to 100×2^8 , $n \in \{100, 200, 400, 800, 1600, 3200, 6400, 12800, 25600\}$. In this experiment, we randomly create an input candidate links set which the size is n , and using the size of output set to evaluate the performance of LSDA. The simulation present that the size of the maximum independent set grows along with the increase of the number of the request link, but due to

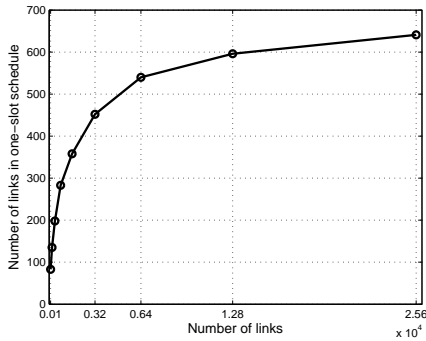


Fig. 3. LSDA Scheduling Algorithm. Nodes random distribution, $\theta = 120^\circ$, the number of links in one-slot schedule show the perform of the algorithm.

the limitations of the space, growth rate becomes slower and tends to balance.

We have a general impression to the performance of LSDA algorithm from above result. Then Now we give some results from other aspect to the algorithm. First, the influence of directional angle, the range of the angle satisfied $\theta \in \{30^\circ, 60^\circ, 90^\circ, \dots, 360^\circ\}$, and we make a comparison experiment $n = 100$ and $n = 3200$. In Fig. 4. We known that the angle make a great influence to the performance of the LSDA scheduling algorithm, when other parameters were fixed. The smaller the angle, the better performance of the algorithm. In the case of maximum angle $\theta = 360^\circ$, the algorithm has a worst performance.

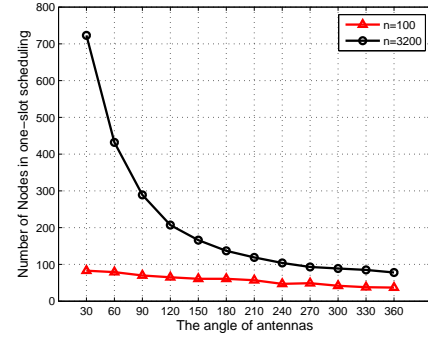


Fig. 4. Influence of Directional Angle. Change the angle of the directional antennas, compared the performance in high density with low density.

Figure 5 show the results of the influence to the algorithm by the antenna gain. We set up a fixed directional angle $\theta = 60^\circ$, antenna gain $G \in \{5, 10, 15, 20, 25, 30, 35, 40\}$, adopt double groups $n = 100$, $n = 3200$. According to the formula of SINR, we known that the changes of the antennas gain have a small effect to the result of algorithm. The simulation results shown that even if there have some fluctuation, the antennas gain have limited impact to the algorithm.

The above results were just get by separate analyzed of the OSML algorithm. Now we do some comparison between OSML algorithm and the one-slot scheduling algorithm. Parameter settings: $Gain = 20$, $\theta = 60^\circ$, and the number of links $n \in \{100, 200, 400, 800, 1600, 3200, 6400, 12800, 25600\}$. Shown in Fig. 6, in random network topology, the performance of the OSML algorithm is batter than the one-slot algorithm. This is advantage of directional antennas, reduce the interference range by every nodes. In a fixed range of scene, more link can be allow for transmit together. In Fig. 7, we change the angle of the antenna $\theta = 360^\circ$, two curves of the algorithm are very similar in the condition of lower density, but slightly worse than one-slot algorithm in high density network. This performance is the comparison of the algorithm. OSML was designed to adapt the situation of direction. So the OSML has limited performance than other omnidirectional algorithm.

To sum up, in random distribution wireless network, it has get an obvious effect to reduce the interference when multiple links transmission simultaneously by using directional anten-

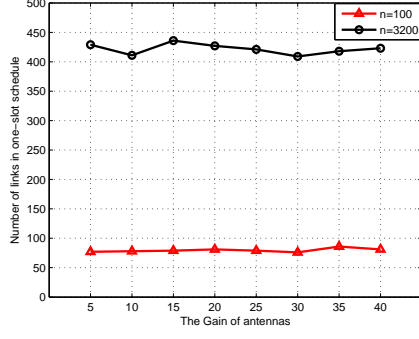


Fig. 5. Influence of Antenna Gain. Change the gain of the directional antennas, compared the performance in high density with low density.

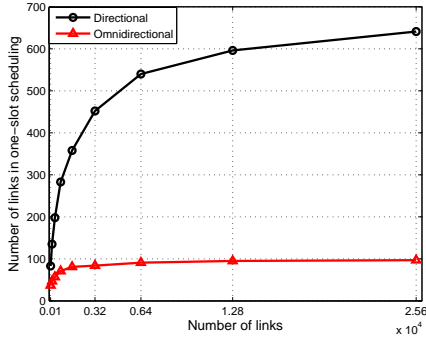


Fig. 6. Comparison between OSML and One-slot Algorithm. Use the same simulation data which generated at randomly by computer, but add the parameter of directional angle $\theta = 60^\circ$ or $\theta = 120^\circ$ when use to running OSML algorithm.

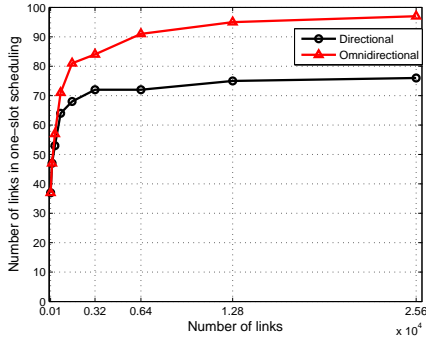


Fig. 7. Comparison between OSML and One-slot Algorithm. Set the directional angle $\theta = 360^\circ$, which means use same random data running in different algorithm, compared the performance of two algorithm.

nas. The LSDA scheduling algorithm have better performance than this omnidirectional algorithm, and we also obtain that the antenna gain have a limited effect to the algorithm, but seriously influence by directional angle.

Now we take full-duplex transmission into consideration, the default configuration parameter were list in Table I.

When take full-duplex into consideration, we give our performance result shown in Figure 8 and Figure 9.

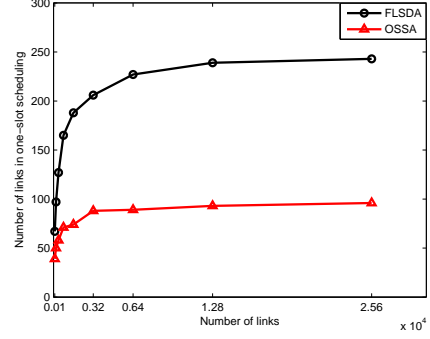


Fig. 8. Comparison between FLSDA and OOSA. Set the directional angle $\theta = 120^\circ$.

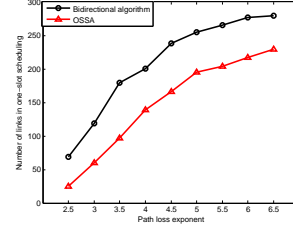


Fig. 9. Comparison between FLSDA and OOSA. Set the directional angle $\theta = 120^\circ$.

Figure 8, we compared our full-duplex algorithm FLSDA with OOSA while set number of links $n \in \{100, 200, 400, 800, 1600, 3200, 6400, 12800, 25600\}$, and the angle of directional antenna is set as $\theta = 120^\circ$. Its obvious that two curves of the algorithm perform very close to each other when the density of the links is low. However the FLSDA behaves significantly better than OOSA as the number of links increases and both of them tend to a fixed number and keep steady.

Next, we analyze the influence of the path loss exponent α on the results (Figure 9). In this simulation, we make comparison between the performance of FLSDA and OOSA while set the directional angle $\theta = 120^\circ$ and the link numbers $n = 400$. It can be seen that the performance of FLSDA is superior to OOSA, even though the performance change of two algorithms influence by path loss exponent is similar. Both of the results of two algorithm keep increasing with the increase

of the path loss exponent but our algorithm also performs better than OSSA.

VI. CONCLUSIONS

In this paper, we developed a directional interference model for wireless network where the nodes using directional antennas. Based on this directional interference model, we proposed a LSDA and FLSDA approximation algorithm for the half-duplex and full-duplex scenarios of links scheduling, respectively. We give simulation and mathematical analysis to show the performance of algorithm. The results of experiments proved that the performance of LSDA was greatly affected by antenna interference angle, but not sensitive to antenna gains. Compared with the omni-directional antennas algorithm, LSDA and FLSDA demonstrate the directional antenna has great advantage in the field of links scheduling. However, there are several challenges to LSDA. The radiation beam of directional antenna is more complicated than we thought.

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