Abstract—Previous work on topology control usually assumes homogeneous wireless nodes with uniform transmission ranges. In this paper, we propose two localized topology control algorithms for heterogeneous wireless multi-hop networks with non-uniform transmission ranges: Directed Relative Neighborhood Graph (DRNG) and Directed Local Spanning Subgraph (DLSS). In both algorithms, each node selects a set of neighbors based on the locally collected information. We prove that (1) the topologies derived under DRNG and DLSS preserve the network connectivity; (2) the out degree of any node in the resulting topology by DLSS is bounded, while the out degree cannot be bounded in DRNG; and (3) the topologies generated by DRNG and DLSS preserve the network bi-directionality.

I. INTRODUCTION

Energy efficiency [1] and network capacity are perhaps two of the most important issues in wireless ad hoc networks and sensor networks. Topology control algorithms have been proposed to maintain network connectivity while reducing energy consumption and improving network capacity. The key idea to topology control is that, instead of transmitting using the maximal power, nodes in a wireless multi-hop network collaboratively determine their transmission power and define the network topology by forming the proper neighbor relation under certain criteria.

By enabling wireless nodes to use adequate transmission power (which is usually much smaller than the maximal transmission power), topology control can not only save energy and prolong network lifetime, but also improve spatial reuse (and hence the network capacity) [2] and mitigate the MAC-level medium contention [3]. Several topology control algorithms [3][10] have been proposed to create power-efficient network topology in wireless multi-hop networks with limited mobility (a summary is given in Section III). However, most of them assume homogeneous wireless nodes with uniform transmission ranges (except [4]).

The assumption of homogeneous nodes does not always hold in practice, since even devices of the same type may have slightly different maximal transmission power. There also exist heterogeneous wireless networks in which devices have dramatically different capabilities, for instance, the communication network in the Future Combat System which involves wireless devices on soldiers, vehicles and UAVs. As will be exemplified in Section III, most existing algorithms cannot be directly applied to heterogeneous wireless multi-hop networks in which the transmission range of each node may be different.

To the best of our knowledge, this paper is the first effort to address the connectivity and bi-directionality issue in the heterogeneous wireless networks.

In this paper, we propose two localized topology control algorithms for heterogeneous wireless multi-hop networks with non-uniform transmission ranges: Directed Relative Neighborhood Graph (DRNG) and Directed Local Spanning Subgraph (DLSS). In both algorithms, the topology is constructed by having each node build its neighbor set and adjust its transmission power based on the locally collected information.

We are able to prove that (1) the topology derived under both DRNG and DLSS preserves network connectivity, i.e., if the original topology generated by having every node use its maximal transmission power is strongly connected, then the topologies generated by both DRNG and DLSS are also strongly connected; (2) the out degree of any node in the topology by DLSS is bounded, while the out degree cannot be bounded in DRNG; and (3) the topologies generated by DRNG and DLSS preserve the network bi-directionality, i.e., if the original topology by having every node use its maximal transmission power is bi-directional, then the topology generated by either DRNG or DLSS is also bi-directional after some simple operations.

Simulation results indicate that, compared with the other known topology control algorithms that can be applied to heterogeneous networks, DRNG and DLSS have smaller average node degree (both logical and physical) and smaller average link length. The former reduces the MAC-level contention, while the latter implies a small transmission power needed to maintain connectivity.

The rest of the paper is organized as follows. In Section II, we give the network model. In Section III, we summarize previous work on topology control, and give examples to show why existing algorithms cannot be directly applied to heterogeneous networks. Following that, we present both the DRNG and DLSS algorithms in Section IV, and prove several of their useful properties in Section V. Finally, we evaluate the performance of the proposed algorithms in Section VI, and conclude the paper in Section VII.
This weight function ensures that two edges with different end-vertices have different weights. Note, however, that \( w(u, v) = w(v, u) \).

**Definition 3 (Neighbor Set):** Node \( v \) is a neighbor of node \( u \) under an algorithm \( A \), denoted \( u \rightarrow^A v \), if and only if there exists an edge \((u, v)\) in the topology generated by the algorithm. In particular, we use \( u \rightarrow v \) to denote the neighbor relation in \( G \). \( u \rightarrow^A v \) if and only if \( u \rightarrow v \) and \( v \rightarrow u \). The Neighbor Set of node \( u \) is \( N^A_u = \{ v \in V(G) : u \rightarrow^A v \} \).

II. NETWORK MODEL

Consider a set of nodes(vertices), \( V = \{v_1, v_2, \ldots, v_n\} \), which are randomly distributed in the 2-D plane. Assume the area that a transmission can cover is a disk. We define the range of a node \( v_i \) as the radius of the disk that \( v_i \) can cover using its maximal transmission power, denoted \( r_{v_i} \). In a heterogeneous network, the transmission ranges of all nodes may not be the same. Let \( r_{min} = \min_{v \in V} \{r_v\} \) and \( r_{max} = \max_{v \in V} \{r_v\} \).

We denote the network topology generated by having each node use its own maximal transmission power as a simple directed graph \( G = (V(G), E(G)) \), where \( E(G) = \{ (u, v) : d(u, v) \leq r_u, u, v \in V(G) \} \) is the edge(link) set of \( G \), and \( d(u, v) \) is the Euclidean distance between node \( u \) and node \( v \). Note that \((u, v)\) is an ordered pair representing an edge from node \( u \) to node \( v \), i.e., \((u, v)\) and \((v, u)\) are two different edges. A unique \( id \) (such as an IP/MAC address) is assigned to each node. Here we let \( id(u_i) = i \) for simplicity.

We assume that the wireless channel is symmetric and obstacle-free, and each node is equipped with the capability to gather its location information via, for example, GPS for outdoor applications and pseudolite [11] for indoor applications, and many other lightweight localization techniques for wireless networks (see [12] for a summary).

Before delving into the technical discussion and algorithm description, we give the definition of several terms that will be used throughout the paper.

**Definition 1 (Reachable Neighborhood):** The reachable neighborhood \( N^R_u \) is the set of nodes that node \( u \) can reach using its maximal transmission power, i.e., \( N^R_u = \{ v \in V(G) : d(u, v) \leq r_u \} \). For each node \( u \in V(G) \), let \( G^R_u = (V(G^R_u), E(G^R_u)) \) be an induced subgraph of \( G \) such that \( V(G^R_u) = N^R_u \).

**Definition 2 (Weight Function):** Given two edges \((u_1, v_1), (u_2, v_2) \in E \) and the Euclidean distance function \( d(\cdot, \cdot) \), weight function \( w : E \mapsto R \) satisfies:

\[
\begin{align*}
&w(u_1, v_1) > w(u_2, v_2) \\
\iff & d(u_1, v_1) > d(u_2, v_2) \\
\text{or} & \left(d(u_1, v_1) = d(u_2, v_2) \\
& \land \max(id(u_1), id(v_1)) > \max(id(u_2), id(v_2))\right) \\
\text{or} & \left(d(u_1, v_1) = d(u_2, v_2) \\
& \land \max(id(u_1), id(v_1)) = \max(id(u_2), id(v_2)) \\
& \land \min(id(u_1), id(v_1)) > \min(id(u_2), id(v_2))\right). 
\end{align*}
\]

This weight function ensures that two edges with different end-vertices have different weights. Note, however, that \( w(u, v) = w(v, u) \).

**Definition 4 (Topology):** The topology generated by an algorithm \( A \) is a directed graph \( G_A = (E(G_A), V(G_A)) \), where \( V(G_A) = V(G), E(G_A) = \{ (u, v) \in E(G) : u \rightarrow^A v \} \).

**Definition 5 (Radius):** The radius, \( R_u \), of node \( u \) is defined as the distance between node \( u \) and its farthest neighbor (in terms of Euclidean distance), i.e., \( R_u = \max_{v \in N_A(u)} \{ d(u, v) \} \).

**Definition 6 (Connectivity):** For any topology generated by an algorithm \( A \), node \( u \) is said to be connected to node \( v \) (denoted \( u \Rightarrow v \)) if there exists a path \((p_0 = u, p_1, \ldots, p_{m-1}, v = v)\) such that \( p_i \rightarrow^A p_{i+1}, i = 0, 1, \ldots, m - 1 \), where \( pk \in V(G_A), k = 0, 1, \ldots, m \). It follows that \( u \Rightarrow v \) if \( u \Rightarrow p \) and \( p \Rightarrow v \) for some \( p \in V(G_A) \).

**Definition 7 (Bi-Directionality):** A topology generated by an algorithm \( A \) is bi-directional, if for any two nodes \( u, v \in V(G_A), u \in N_A(v) \) implies \( v \in N_A(u) \). In other words, the topology generated by \( A \) is bi-directional if all edges in the topology are bi-directional.

**Definition 8 (Bi-Directional Connectivity):** For any topology generated by an algorithm \( A \), node \( u \) is said to be bi-directionally connected to node \( v \) (denoted \( u \Rightarrow^B v \)) if there exists a path \((p_0 = u, p_1, \ldots, p_{m-1}, v = v)\) such that \( p_i \Rightarrow^B p_{i+1}, i = 0, 1, \ldots, m - 1 \), where \( pk \in V(G_A), k = 0, 1, \ldots, m \). It follows that \( u \Rightarrow^B v \) if \( u \Rightarrow^B p \) and \( p \Rightarrow v \) for some \( p \in V(G_A) \).

Deriving network topology consisting of only bi-directional links facilitates link level acknowledgment, which is a critical operation for packet transmissions and retransmissions over unreliable wireless media. Bi-directionality is also important in floor acquisition mechanisms such as the RTS/CTS mechanism in IEEE 802.11.

**Definition 9 (Addition and Removal):** The operation Addition is to add an extra edge \((v, u)\) into \( G_A \) if \((u, v) \in E(G_A), (v, u) \notin E(G_A)\), and \( d(u, v) \leq r_u \). The operation Removal is to delete any edge \((v, u)\) in \( E(G_A) \) if \((v, u) \notin E(G_A)\). Let \( G^A_A \) and \( G^A_A \) denote the resulting topologies after applying Addition and Removal to \( G_A \), respectively.

Both the Addition and Removal operations attempt to create a bi-directional topology by removing uni-directional edges or converting uni-directional edges into bi-directional. The resulting topology after Removal is always bi-directional, although it may not be strongly connected. The resulting topology after Addition is not necessarily bi-directional, as it essentially tries to increase the transmission power of a node \( v \) to a level that may be beyond its capability.

III. RELATED WORK AND WHY THEY CANNOT BE DIRECTLY APPLIED TO HETEROGENEOUS NETWORKS

Several topology control algorithms [3]–[10] have been proposed. In this section, we first summarize these algorithms and then give examples on why they cannot be directly applied to heterogeneous networks.

A. Related Work

Rodoplu et al. [4] (denoted R&M) introduced the notion of relay region and enclosure for the purpose of power control. Instead of transmitting directly, a node chooses to...
relay through other nodes if less power will be consumed. It is shown in the paper that the network is strongly connected if every node maintains links with the nodes in its enclosure and the resulting topology is a minimum power topology. The major drawback is that it requires an explicit propagation channel model to compute the relay region (in the simulation study presented in Section VI, we assume that the free-space model is used), hence the resulting topology is sensitive to the model used in the computation. Also, it assumes there is only one data sink (destination) in the network.

Ramanathan et al. [5] presented two centralized algorithms to minimize the maximal power used per node while maintaining the (bi)connectivity of the network. They introduced two distributed heuristics for mobile networks. Both centralized algorithms require global information, and thus cannot be directly deployed in the case of mobility. On the other hand, the proposed heuristics cannot guarantee the preservation of the network connectivity.

COMPOW [3] and CLUSTERPOW [7] are approaches implemented in the network layer. Both hinge on the idea that if each node uses the smallest common power required to maintain network connectivity, the traffic carrying capacity of the entire network is maximized, the battery life is extended, and the MAC-level contention is mitigated. The major drawback is its significant message overhead, since each node has to run multiple daemons, each of which has to exchange link state information with their counterparts at other nodes.

CBTC(α) [6] is a two-phase algorithm in which each node finds the minimum power p such that some node can be reached in every cone of degree α. The algorithm has been proved to preserve network connectivity if α < 5π/6. Several optimization methods (that are applied after the topology is derived under the base algorithm) are also discussed to further reduce the transmitting power.

To facilitate the following discussion, the definition of the Relative Neighborhood Graph (RNG) is given below.

Definition 10 (Neighbor Relation in RNG): For RNG [13], [14], \( u \xrightarrow{\text{RNG}} v \) if and only if there does not exist a third node \( p \) such that \( w(u, p) < w(u, v) \) and \( w(p, v) < w(u, v) \). Or equivalently, there is no node inside the shaded area in Fig. 1(a).

Borbash and Jennings [8] proposed to use RNG for the topology initialization of wireless networks. Based on the local knowledge, each node makes decisions to derive the network topology based on RNG. The network topology thus derived has been reported to exhibit good overall performance in terms of power usage, low interference, and reliability.

Li et al. [9] presented the Localized Delaunay Triangulation, a localized protocol that constructs a planar spanner of the Unit Disk Graph (UDG). The topology contains all edges that are both in the unit-disk graph and the Delaunay triangulation of all nodes. It is proved that the shortest path in this topology between any two nodes \( u \) and \( v \) is at most a constant factor of the shortest path connecting \( u \) and \( v \) in UDG. However, the notion of UDG and Delaunay triangulation cannot be directly extended to heterogeneous networks.

In [10], we proposed LMST (Local Minimum Spanning Tree) for topology control in homogeneous wireless multihop networks. In this algorithm, each node builds its local minimum spanning tree independently and only keeps on-tree nodes that are one-hop away as its neighbors in the final topology. It is proved that (1) the topology derived under LMST preserves the network connectivity; (2) the node degree of any node in the resulting topology is bounded by 6; and (3) the topology can be transformed into one with bidirectional links (without impairing the network connectivity) after removal of all uni-directional links. Simulation results show that LMST can increase the network capacity as well as reduce the energy consumption.

Instead of adjusting the transmission power of individual devices, there also exist other approaches to generate power-efficient topology. By following a probabilistic approach, Santi et al. derived the suitable common transmission range which preserves network connectivity, and established the lower and upper bounds on the probability of connectedness [15]. In [16], a “backbone protocol” is proposed to manage large wireless ad hoc networks, in which a small subset of nodes is selected to construct the backbone. In [17], a method of calculating the power-aware connected dominating sets was proposed to
Fig. 2. An example that shows CBTC \((2, \pi)\) may render disconnectivity in heterogeneous networks. There is no path from \(v_1\) to \(v_3\) due to the loss of edge \((v_2, v_3)\), which is discarded by \(v_2\) since \(v_1\) and \(v_4\) have already provided the necessary coverage.

Fig. 3. An example that shows RNG may render disconnectivity in heterogeneous networks. There is no path from \(v_5\) to \(v_2\) due to the loss of edge \((v_4, v_2)\), which is discarded since \(|(v_4, v_5)| < |(v_4, v_2)|\) and \(|(v_2, v_5)| < |(v_2, v_4)|\).

Fig. 4. An example that shows MRNG may render disconnectivity in heterogeneous networks. There is no path from \(v_3\) to \(v_5\) due to the loss of edge \((v_2, v_5)\), which is discarded since \(|(v_2, v_3)| < |(v_2, v_5)|\) and \(|(v_5, v_3)| < |(v_2, v_5)|\).

Fig. 5. An example that shows the algorithm in which each node builds a local directed minimum spanning tree and only keeps the one-hop neighbors may result in disconnectivity.
establish an underlying topology for the network.

B. Why Existing Algorithms Cannot be Directly Applied to Heterogeneous Networks

Most existing topology control algorithms (except [4]) assume homogeneous wireless nodes with uniform transmission ranges. When directly applied to heterogeneous networks, these algorithms may render disconnectivity. In this subsection, we give several examples to motivate the need for new topology control algorithms for heterogeneous networks.

As shown in Fig. 2 (a)-(b) (note that in Figs. 2–5 we use arrows to indicate the direction of the links to represent a link from $u$ to $v$), the network topology derived under $CBTC(\frac{2}{3} \pi)$ (without optimization) may not preserve the connectivity, when the algorithm is directly applied to a heterogeneous network. $CBTC(\frac{2}{3} \pi)$ also has the same problem.

Similarly we show in Fig. 3 (a)-(b) that the network topology derived under RNG may be disconnected when the algorithm is directly applied to a heterogeneous network. As RNG is defined for undirected graphs, one may tailor the definition of RNG for directed graphs.

Definition 11 (Neighbor Relation in MRNG): For Modified Relative Neighborhood Graph (MRNG), $u \xrightarrow{MRNG} v$ if and only if there does not exist a third node $p$ such that $w(u, p) < w(u, v)$, $d(u, p) \leq r_u$ and $w(p, v) < w(u, v)$, $d(v, p) \leq r_v$ (Fig. 1(b)).

As shown in Fig. 4 (a)-(b), the topology derived under MRNG may still be disconnected (we will give another variation of RNG for directed graphs in the next section).

One possible extension of LMST [10] is for each node to build a local directed minimum spanning tree [18]–[20] and keep only neighbors within one hop. Unfortunately, the resulting topology does not preserve the strong connectivity, as shown in Fig. 5. In the next section, we will improve on this approach to preserve the connectivity.

IV. DRNG AND DLSS

In this section, we propose two localized topology control algorithms for heterogeneous wireless multi-hop networks with non-uniform transmission ranges: Directed Relative Neighborhood Graph (DRNG) and Directed Local Spanning Subgraph (DLSS). In both algorithms, the topology is derived by having each node build its neighbor set and adjust its transmission power based on locally collected information. Several nice properties of both algorithms will be discussed in Section V.

Both algorithms are composed of three phases:

1) Information Collection: each node collects the local information of neighbors such as position and $id$, and identifies the Reachable Neighborhood $N^R_u$.

2) Topology Construction: each node defines (in compliance with the algorithm) the proper list of neighbors for the final topology using the information in $N^R_u$.

3) Construction of Topology with Only Bi-Directional Links (Optional): each node adjusts its list of neighbors to make sure that all the edges are bi-directional.

A. Information collection

The information needed by each node $u$ for topology control is the information of its reachable neighborhood $N^R_u$. This can be obtained locally, in the case of homogeneous networks, by having each node broadcast periodically a Hello message using its maximal transmission power. The information contained in a Hello message should at least include the node $id$ and the position of the node. These periodic messages can be sent either in the data channel or in a separate control channel. In heterogeneous networks, having each node broadcast a Hello message using its maximal transmission power may be insufficient. For example, as shown in Fig. 6, $v_1$ is unable to know the position of $v_4$ since $v_4$ cannot reach $v_1$. We will treat this issue rigorously in Section V-D. For the time being, we assume that by the end of the first phase every node $u$ obtains its $N^R_u$.

B. Topology construction

First we define the neighbor relation used in both algorithms.

Definition 12 (Neighbor Relation in DRNG): For Directed Relative Neighborhood Graph (DRNG), $u \xrightarrow{DRNG} v$ if and only if $d(u, v) \leq r_u$ and there does not exist a third node $p$ such that $w(u, p) < w(u, v)$ and $w(p, v) < w(u, v)$, $d(p, v) \leq r_p$ (see Fig. 1(c)).

Definition 13 (Neighbor Relation in DLSS): For Directed Local Spanning Subgraph (DLSS), $u \xrightarrow{DLSS} v$ if and only if $(u, v) \in E(T_u)$, where $T_u$ is obtained by applying Algorithm 1 to $G^R_u$. $T_u$ is a directed local spanning subgraph that spans $N^R_u$. Hence node $v$ is a neighbor of node $u$ if and only if node $v$ is on node $u$’s directed local MST $T_u$, and is one-hop away from node $u$.

DLSS is a natural extension of LMST [10] for heterogeneous networks. Instead of computing a directed local MST (which minimizes the total cost of the all edges in the subgraph, and is shown to be wrong in Section III-B),
**Algorithm 1 DLSS(u)**

**INPUT:** $G^R_u$, the induced subgraph of $G$ that spans the reachable neighborhood of $u$;  
**OUTPUT:** $T_u = (V_{T_u}, E_{T_u})$, a local spanning subgraph of $G^R_u$;  
1: $V_{T_u} := V$, $E_{T_u} := \emptyset$;  
2: sort all edges in $E(G^R_u)$ in the ascending order of weight (as defined in Definition 2);  
3: for each edge $(u, v)$ in the order do  
4: if $u$ is not connected to $v$ in $T_u$ then  
5: $E_{T_u} := E_{T_u} \cup \{(u, v)\}$;  
6: end if  
7: if $u$ is connected to all nodes in $N_u \subseteq V$ then  
8: exit;  
9: end if  
10: end for

Each node $u$ computes a directed local subgraph according to Algorithm 1 (which minimizes the maximum cost of all edges in the subgraph) and takes on-tree nodes that are one-hop away as its neighbors. 

Each node can broadcast its own maximal transmission power in the Hello message. By measuring the receiving power of Hello messages, each node $u$ can determine the specific power level required to reach each of its neighbors [10]. Node $u$ then uses the power level that can reach its farthest neighbor as its transmission power. This process can be applied without knowing the actual propagation model.

**C. Construction of topology with only bi-directional edges**

As illustrated in the previous section, some links in $G_{DLSS}$ may be uni-directional. There also exist uni-directional links in $G_{DRNG}$. We can apply either Addition or Removal to $G_{DLSS}$ and $G_{DRNG}$ to obtain bi-directional topologies. We will discuss some properties of these solutions in Section V-B.

**V. PROPERTIES OF DRNG AND DLSS**

In this section, we discuss the connectivity, bi-directionality and degree bound of DLSS and DRNG. We always assume $G$ is strongly connected, i.e., $u \Rightarrow v$ in $G$ for any $u, v \in V(G)$.

**A. Connectivity**

**Lemma 1:** For any edge $(u, v) \in E(G)$, we have $u \Rightarrow v$ in $G_{DLSS}$.

**Proof:** Let all the edges $(u, v) \in E(G)$ be sorted in the increasing order of weight, i.e., $w(u_1, v_1) < w(u_2, v_2) < \ldots < w(u_l, v_l)$, where $l$ is the total number. We prove by induction.

1) **Basis:** The first edge $(u_1, v_1)$ satisfies $w(u_1, v_1) = \min_{(u, v) \in E(G)} \{w(u, v)\}$. According to Algorithm 1, $(u_1, v_1)$ and $(v_1, u_1)$ will be inserted into $G_{DLSS}$, i.e., $u_1 \leftrightarrow v_1$.

2) **Induction:** Assume the hypothesis holds for all edges $(u_i, v_i), 1 \leq i < k$, we prove $u_k \Rightarrow v_k$ in $G_{DLSS}$. If $u_k \leftrightarrow v_k$, then $u_k \Rightarrow v_k$. Otherwise in the local topology construction of $u$, before edge $(u_k, v_k)$ was inserted into $T_{u_k}$, there must already exist a path $p = (w_0 = u_k, w_1, w_2, \ldots, w_{m-1}, w_m = v_k)$ from $u_k$ to $v_k$, where $(w_i, w_{i+1}) \in E(T_{u_k}), i = 0, 1, \ldots, m - 1$. Since edges are inserted in a ascending order of weight, we have $w_i < w_{i+1}$. This implies the induction hypothesis to each pair $w_i, w_{i+1}, i = 0, 1, \ldots, m - 1$. If we have $w_i \Rightarrow w_{i+1}$, then the proof is complete. Therefore, we have $w_i \Rightarrow w_{i+1}$, therefore, $u_k \Rightarrow v_k$.

**Theorem 1:** $G_{DLSS}$ preserves the connectivity of $G$, i.e., $G_{DLSS}$ is strongly connected if $G$ is strongly connected.

**Proof:** Suppose $G$ is strongly connected. For any two nodes $u, v \in V(G)$, there exists at least one path $(w_0 = u, w_1, w_2, \ldots, w_m = v)$ from $u$ to $v$, where $(w_i, w_{i+1}) \in E(G), i = 0, 1, \ldots, m - 1$. Since $w_i \Rightarrow w_{i+1}$ by Lemma 1, we have $u \Rightarrow v$.

**Lemma 2:** Given three nodes $u, v, w \in V(G_{DLSS})$ satisfying $w(u, v) > w(u, w)$ and $w(u, v) > w(v, w), d(w, v) \leq r_u$, then $u \Rightarrow v$ in $G_{DLSS}$.

**Proof:** We only need to consider the case where $d(u, v) \leq r_u$, since $d(u, v) > r_u$ would imply $u \Rightarrow v$. Consider the local topology construction of $u$. Before we insert $(u, v)$ into $T_u$, the two edges $(u, p)$ and $(p, v)$ have already been processed since $w(u, p) < w(u, v)$ and $w(p, v) < w(u, v)$. Thus $u \Rightarrow p$ and $p \Rightarrow v$, which means $u \Rightarrow v$. Therefore, $(u, v)$ should not be inserted into $T_u$ according to Algorithm 1, i.e., $u \Rightarrow v$ in $G_{DLSS}$.

**Theorem 2:** The edge set of $G_{DLSS}$ is a subset of the edge set of $G_{DRNG}$, i.e., $E(G_{DLSS}) \subseteq E(G_{DRNG})$.

**Proof:** We prove by contradiction. Given any edge $(u, v) \in E(G_{DLSS})$, assume $(u, v) \notin E(G_{DRNG})$. According to the definition of $DRNG$, there must exist a third node $p$ such that $w(u, p) < w(u, v), d(u, p) \leq r_u$ and $w(p, v) < w(u, v), d(p, v) \leq r_p$. By Lemma 2, $u \Rightarrow v$ in $G_{DLSS}$, i.e., $(u, v) \notin E(G_{DLSS})$.

**Theorem 3 (Connectivity of DRNG):** If $G$ is strongly connected, then $G_{DRNG}$ is also strongly connected.

**Proof:** This is a direct result of Theorem 1 and Theorem 2.

**B. Bi-directionality**

Now we discuss the bi-directionality property of DLSS and DRNG. Since Addition may not always result in bi-directional topologies, we first apply Removal to topologies by DLSS and DRNG. It turns out the simple Removal operation may lead to disconnectivity. Examples are given in Figs. 7–8 to show, respectively, that DLSS and DRNG with Removal may result in disconnectivity.

In general, $G$ may not be bi-directional if the transmission ranges are non-uniform. Since the maximal transmission range can not be increased, it may be impossible to find a bi-directional connected subgraph of $G$ for some cases. An
Fig. 7. An example that shows DLSS with Removal may result in disconnectivity.

Fig. 8. An example that shows DRNG with Removal may result in disconnectivity.

Fig. 9. The definition of $\text{Cone}(u, \alpha, v)$.
Therefore, node $u$ is at most 6.

**Proof:** Let $N(u)$ be the set of neighbors of $u$ in $G_{DLSS}$ that are inside $\text{Disk}(u, r_{\text{min}})$. Let the nodes in $N(u)$ be ordered such that for the $i$th node $w_i$ and the $j$th node $w_j$ ($j > i$), $w(u, w_j) > w(u, w_i)$. By Lemma 2, we have $w(u, w_j) \leq w(w_i, w_j)$ (otherwise $u \rightarrow w_j$). Thus $\angle w_i w u \angle w_j \geq \pi/3$, i.e., node $w_j$ cannot reside inside $\text{Cone}(u, 2\pi/3, w_i)$. Therefore, node $u$ cannot have neighbors other than node $w_i$ inside $\text{Cone}(u, 2\pi/3, w_i)$. By induction on the rank of nodes in $N(u)$, the maximal number of neighbors that $u$ can have is at most 6.

**Theorem 6:** The out degree of node in $G_{DLSS}$ is bounded by a constant that depends only on $r_{\text{max}}$ and $r_{\text{min}}$.

**Proof:** For any node $u$ in $G_{DLSS}$, there are at most 6 neighbors inside $\text{Disk}(u, r_{\text{min}})$ from Theorem 5. Also from Corollary 1, the set of disks $\{\text{Disk}(v, r_{\text{max}}') : v \in N_{DLSS}(u), v \notin \text{Disk}(u, r_{\text{min}})\}$ are disjoint. Therefore, the total number of neighbors of $u$ is bounded by:

$$c_1 = 6 + \left\lceil \frac{\pi (r_{\text{max}} + \frac{r_{\text{min}}}{2})^2 - (\frac{r_{\text{min}}}{2})^2}{\pi (r_{\text{min}}/2)^2} \right\rceil = 4[\beta(\beta+1)] + 6,$$

where $\beta = \frac{r_{\text{max}}}{r_{\text{min}}}$. Actually we can observe that Fig. 10 shows the scenario where the maximum out degree of $u$ is achieved if $\epsilon \to 0$. Therefore, we can further tighten the bound. Since the hexagonal area (as shown in Fig. 10) centered at every neighbor of $u$ is disjoint with each other, the total number of neighbors of $u$ is bounded by:

$$c_2 = \left\lceil \frac{\pi r_{\text{max}} + \frac{r_{\text{min}}}{2}}{2 \frac{\sqrt{3}}{2} r_{\text{min}}} \right\rceil - 1 - \left\lceil \frac{2\pi}{\sqrt{3}} (\beta + \frac{1}{\sqrt{3}})^2 \right\rceil - 1.$$
$r_u' \leq r_u$ since for any $v \in N_u^{R'}, d(u,v) \leq r_u$. Let $r_{\min}' = \min_{v \in V} \{r_v'\}$ and $r_{\max}' = \max_{v \in V} \{r_v'\}$. By requiring each node $u$ to broadcast its position and id to all other nodes within $r_u$, we are able to determine $N_u^{R'}$ and $r_u'$. We can then apply DRNG and DLSS on top of $G'$ and prove that Theorems 1-5 still hold even if the original topology is $G'$.

**Theorem 7:** Theorems 1–6 still holds if the original topology is $G'$.

**Proof:** We replace $G$, $r_u$, $N_u^{R}$, $r_{\min}$, and $r_{\max}$ with $G'$, $r_u'$, $N_u^{R'}$, $r_{\min}'$ and $r_{\max}'$ in the proof of Lemma 1–2 and Theorem 1–6. Then following the same line of arguments, we can prove that they still hold if the original topology is $G'$.

**Theorem 8:** If the original topology is $G'$ (which is a subgraph of $G$), $G_{DLSS}$ and $G_{DRNG}$ are bi-directional after Addition or Removal.

**Proof:** We apply Theorem 4 to $G'$, for $G'$ is bi-directional.
VI. SIMULATION STUDY

In this section, we evaluate the performance of R&M, DRNG, and DLSS by simulations. All three algorithms are known to preserve network connectivity in heterogeneous networks.

In the first simulation, 50 nodes are uniformly distributed in a $1000m \times 1000m$ region. The transmission ranges of nodes are uniformly distributed in $[200m, 250m]$. Fig. 12 gives the topologies derived using the maximal transmission power (labeled as NONE), R&M (under the two-ray ground model), DRNG, and DLSS for one simulation instance. As shown in Fig. 12, R&M, DRNG and LMST all significantly reduce the average node degree, while maintaining network connectivity. Moreover, both DRNG and DLSS outperforms R&M in the sense that fewer edges are formed in the topology.

Fig. 13 shows the average radius and the average link length for the topologies derived under NONE (no topology control), R&M, DRNG, and DLSS. DLSS outperforms the others, which implies that DLSS can provide a better spatial reuse and use less energy to communicate.

We also compare the out degree of the topologies by different algorithms. The result of NONE is not shown because its out degrees increase almost linearly with the number of nodes and are significantly larger than those under R&M, DRNG, and DLSS. DLSS outperforms the others, which implies that DLSS can provide a better spatial reuse and use less energy to communicate.

We also compare the out degree of the topologies by different algorithms. The result of NONE is not shown because its out degrees increase almost linearly with the number of nodes and are significantly larger than those under R&M, DRNG, and DLSS. DLSS outperforms the others, which implies that DLSS can provide a better spatial reuse and use less energy to communicate.

In the second simulation, we vary the number of nodes in the region from 100 to 300, and each data point is an average of 50 simulation runs. The transmission ranges of nodes are uniformly distributed in $[200m, 250m]$. Fig. 13 shows the average radius and the average link length for the topologies derived under NONE (no topology control), R&M, DRNG, and DLSS. DLSS outperforms the others, which implies that DLSS can provide a better spatial reuse and use less energy to communicate.

We also compare the out degree of the topologies by different algorithms. The result of NONE is not shown because its out degrees increase almost linearly with the number of nodes and are significantly larger than those under R&M, DRNG, and DLSS. DLSS outperforms the others, which implies that DLSS can provide a better spatial reuse and use less energy to communicate.

In Fig. 14 shows the average logical/physical out degree for the topologies derived by R&M, DRNG, and DLSS. The average out degrees under R&M and DRNG increase with the increase in the number of nodes, while those under DLSS actually decrease. Fig. 15 shows the average maximum logical degree and the largest maximum logical
out degree for each number of nodes. The largest maximum logical degree under DLSS is at most 4, and is well below the theoretical upper bound obtained in Theorem 6. Also DLSS has much smaller degrees than the other topologies. Similar results can be observed in Fig. 16 for physical degrees. The only difference is that the physical degrees are in general larger than the logical degrees for the same network.

**VII. CONCLUSIONS**

In this paper, we have proposed two local topology control algorithms, Directed Relative Neighborhood Graph (DRNG) and Directed Local Spanning Subgraph (DLSS), for heterogeneous wireless multi-hop networks in which each node may have different maximal transmission ranges. We show that as most existing topology control algorithms (except R&M [4]) do not consider the fact that nodes may have different maximal transmission ranges, they render disconnected network topol-ogy when directly applied to heterogeneous networks. Then we devise DRNG and DLSS and prove that (i) both DRNG and DLSS preserve network connectivity; (ii) both DRNG and DLSS preserve network bi-directionality if *Addition* and *Remove* operations are applied to the topologies derived under these algorithms; and (iii) the out degree of any node is bounded in the topology derived under DLSS, while that may be unbounded under DRNG. The simulation study validates the superiority of DRNG and DLSS over R&M.

As part of our future research, we will pursue the following open problems: (1) given a topology in which each node transmits with different maximal transmission power, what is the probability that the topology is bi-directional with respect to the distribution and the density of nodes, and the distribution of the transmission ranges? and (2) How will MAC-level interference affect network connectivity and bi-directionality?
REFERENCES


