

Chapter 3

ON THE CONSTRUCTION OF VIRTUAL BACKBONE FOR AD HOC WIRELESS NETWORK

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Abstract Ad hoc wireless network is featured by a dynamic topology. There is no fixed infrastructure as compared with wired network. Every host can move to any direction at any speed. This characteristic puts special challenges in routing protocol design. Most existing well-known routing protocols use flooding for route construction. But, flooding suffers from the notorious *broadcast storm problem* which causes excessive *redundancy*, *contention* and *collision* in the network. One solution to overcome this problem is to compute a virtual backbone based on the physical topology, and run any existing routing protocol over the virtual backbone. In our study, the virtual backbone is approximated by a *minimum*

connected dominating set (MCDS). We propose a distributed algorithm which computes a small CDS. The performance of our algorithm is witnessed by simulation results and theoretical analysis.

Keywords: Multihop ad hoc wireless network, connected dominating set, virtual backbone routing

1. Introduction

Ad hoc wireless network has applications in emergency search-and-rescue operations, decision making in the battlefield, data acquisition operations in inhospitable terrain, etc. It is featured by dynamic topology (infrastructure-less), multihop communication, limited resources (bandwidth, CPU, battery, etc) and limited security. These characteristics put special challenges in routing protocol design.

Existing routing protocols (see [10, 12, 14, 15]) can be classified into three categories: *proactive*, *reactive* and the combination of the two. Proactive routing protocols ask each host to maintain global topology information, thus a route can be provided immediately when requested. Protocols in this category suffer from lower scalability and high protocol overhead. Reactive routing protocols have the feature *on-demand*. Each host computes route for a specific destination only when necessary. Topology changes that do not influence active routes do not trigger any route maintenance function, thus communication overhead is lower compared to proactive routing protocol. The third category maintains partial topology information in some hosts. Routing decisions are made either proactively or reactively. One important observation on these protocols is that none of them can avoid the involvement of *flooding*. For example, proactive protocols rely on flooding for the dissemination of topology update packets, and reactive protocols rely on flooding for route discovery.

Flooding suffers from the notorious *broadcast storm problem* [13]. Broadcast storm problem refers to the fact that flooding may result in excessive *redundancy*, *contention*, and *collision*. This causes high protocol overhead and interference to ongoing traffic. On the other hand, flooding is very *unreliable* [11], which means that *not all* hosts get all the broadcast messages when free from collisions. Sinha et. al. in [16] claimed that “in moderately sparse graphs the expected number of nodes in the network that will receive a broadcast message was shown to be as low as 80%.” In reactive protocols, the unreliability of flooding may obstruct the detection of the shortest path, or simply can’t detect any path at all, even though there exists a path. In proactive protocols, the unreliability of flooding may cause the global topology information to become obsolete, thus causing the newly-computed path obsolescence.

Recently an approach based on overlaying a virtual infrastructure on an ad hoc network was proposed in [16]. Routing protocols are operated over this infrastructure, which is termed *core*. All core hosts form a dominating set. The key feature in this approach is the new *core broadcast mechanism* which uses unicast to replace the flooding mechanism used by most on-demand routing protocols. The unicast of route requests packets to be restricted to core nodes and a (small) subset of non-core nodes. Simulation results when running DSR (Dynamic Source Routing [12]) and AODV (Ad hoc On-demand Distance Vector routing [15]) over the core indicate that the core structure is *effective* in enhancing the performance of the routing protocols. Prior to this work, inspired by the *physical backbone* in a wired network, many researchers proposed the concept of *virtual backbone* for unicast, multicast/broadcast in ad hoc wireless networks (see [2, 8, 17]).

In this chapter, we will study the problem of efficiently constructing virtual backbone for ad hoc wireless networks. The number of hosts forming the virtual backbone must be as small as possible to decrease protocol overhead. The algorithm must be time/message efficient due to resource scarcity. We use a connected dominating set (CDS) to approximate the virtual backbone. We assume a given ad hoc network instance contains n hosts. Each host is in the ground and is mounted by an omni-directional antenna. Thus the transmission range of a host is a disk. We further assume that each transceiver has the same communication range R . Thus the footprint of an ad hoc wireless network is a unit-disk graph. In graph-theoretic terminology, the network topology we are interested in is a graph $G = (V, E)$ where V contains all hosts and E is the set of links. A link between u and v exists if they are separated by the distance of at most R . In a real world ad hoc wireless network, sometimes even when v is located in u 's transmission range, v is not reachable from u due to *hidden/exposed terminal problems*. But in this chapter we only consider bidirectional links. From now on, we use host and node interchangeably to represent a wireless mobile.

There exist several distributed algorithms ([1, 8, 17]) for MCDS computation in the context of ad hoc wireless networking. The one in [1] first builds a rooted tree distributedly. Then the status (inside or outside of the CDS) is assigned for each host based on the level of the host in the tree. Das and Bharghavan [8] provide the distributed implementation of the two centralized algorithms given by Guha and Khuller [9]. Both implementations suffer from high message complexities. The one given by Wu and Li in [17] has no performance analysis, but it needs at least two-hop neighborhood information. The status of each host is assigned based on the connectivity of its neighbors. We will compare our algorithm with the other approaches [1, 8, 17] in Sections 3 and 4.

The remainder of this chapter is organized as follows. Section 2 provides basic concepts related to this topic. Our algorithm and its theoretic performance analysis are presented in Section 3. Simulation results are demonstrated in Section 4. Section 5 concludes the chapter.

2. Preliminaries

Given graph $G = (V, E)$, two vertices are *independent* if they are not neighbors. For any vertex v , the set of *independent neighbors* of v is a subset of v 's neighbors such that any two vertices in this subset are independent. An *independent set (IS)* S of G is a subset of V such that $\forall u, v \in S, (u, v) \notin E$. S is *maximal* if any vertex not in S has a neighbor in S (denoted by MIS).

A *dominating set (DS)* D of G is a subset of V such that any node not in D has at least one neighbor in D . If the induced subgraph of D is connected, then D is a *connected dominating set (CDS)*. Among all CDSs of graph G , the one with minimum cardinality is called a *minimum connected dominating set (MCDS)*. Computing an MCDS in a unit graph is NP-hard [7]. Note that the problem of finding an MCDS in a graph is equivalent to the problem of finding a spanning tree (ST) with maximum number of leaves. All non-leaf nodes in the spanning tree form the MCDS. An MIS is also a DS.

For a graph G , if $e = (u, v) \in E$ iff $length(e) \leq 1$, then G is called a *unit-disk graph*. We will only consider unit-disk graphs in this chapter. From now on, when we say a “graph G ”, we mean a “unit-disk graph G ”. The following lemma was proved in [1]. This lemma relates the size of any MIS of unit-disk graph G to the size of its optimal CDS.

Lemma 2.1. [1] *The size of any MIS of G is at most $4 \times opt + 1$, where opt is the size of any MCDS of G .*

For a minimization problem \mathcal{P} , the *performance ratio* of an approximation algorithm A is defined to be $\rho_A = \sup_{i \in I} \frac{A_i}{opt_i}$, where I is the set of instances of \mathcal{P} , A_i is the output from A for instance i and opt_i is the optimal solution for instance i . In other words, ρ is the supreme of $\frac{A}{opt}$ among all instances of \mathcal{P} .

3. An 8-Approximate Algorithm to Compute CDS

In this section, we propose a distributed algorithm to compute CDS. This algorithm contains two phases. First, we compute a maximal independent set (MIS); then we use a tree to connect all vertices in the MIS. We will show that our algorithm has performance ratio at most 8 and is message and time efficient.

3.1. Algorithm Description

Initially each host is colored *white*. A dominator is colored *black*, while a dominatee is colored *gray*. The *effective degree* of a vertex is the total number of white neighbors. We assume that each vertex knows its distance-one neighbors and their effective degrees d^* . This information can be collected by periodic or event-driven hello messages.

We also designate one host as the *leader*. This is a realistic assumption. For example, the leader can be the commander's mobile for a platoon of soldiers in a mission. If it is impossible to designate any leader, a distributed leader-election algorithm can be applied to find out a leader. This adds message and time complexity. The best leader-election algorithm (see [4]) takes time $O(n)$ and message $O(n \log n)$ and these are the best-achievable results. Assume host s is the leader.

Phase 1. Host s first colors itself black and broadcasts message DOMINATOR.

Any white host u receiving DOMINATOR message the first time from v colors itself gray and broadcasts message DOMINATEE. u selects v as its dominator. A white host receiving at least one DOMINATEE message becomes active. An active white host with highest (d^*, id) among all of its active white neighbors will color itself black and broadcast message DOMINATOR. A white host decreases its effective degree by 1 and broadcasts message DEGREE whenever it receives a DOMINATEE message. Message DEGREE contains the sender's current effective degree. A white vertex receiving a DEGREE message will update its neighborhood information accordingly. Each gray vertex will broadcast message NUMOFBLACKNEIGHBORS when it detects that none of its neighbors is white. Phase 1 terminates when no white vertex left.

Phase 2. When s receives message NUMOFBLACKNEIGHBORS from all of its gray neighbors, it starts phase 2 by broadcasting message M. A host is "ready" to be explored if it has no white neighbors. We will use a tree to connect all black hosts generated in Phase 1. The idea is to pick those gray vertices which connect to many black neighbors. We will modify the classical distributed depth first search spanning tree algorithm given in [3] to compute the tree.

A black vertex without any dominator is *active*. Initially no black vertex has a dominator and all hosts are *unexplored*. Message M contains a field *next* which specifies the next host to be explored. A gray vertex with at least one active black neighbor is *effective*. If M is built by a black vertex, its *next* field contains the *id* of the unexplored gray neighbor which connects to maximum number of active black hosts. If M is built by a gray vertex, its *next* field contains the *id* of any unexplored black neigh-

bor. Any black host u receiving an M message the first time from a gray host v sets its dominator to v by broadcasting message PARENT. When a gray host u receives message M from v that specifies u to be explored next, u then colors itself black, sets its dominator to v and broadcasts its own M message. Any gray vertex receiving message PARENT from a black neighbor will broadcast message NUMOFBLACKNEIGHBORS, which contains the number of active black neighbors. A black vertex becomes *inactive* after its dominator is set. A gray vertex becomes *ineffective* if none of its black neighbors is active. A gray vertex without active black neighbor, or a black vertex without effective gray neighbor, will send message DONE to the host which activates its exploration or to its dominator. When s gets message DONE and it has no effective gray neighbors, the algorithm terminates.

Note that phase 1 sets the dominators for all gray vertices. Phase 2 may modify the dominator of some gray vertex. The main job for phase 2 is to set a dominator for each black vertex. All black vertices form a CDS.

In Phase 1, each host broadcasts each of the messages DOMINATOR and DOMINATEE at most once. The message complexity is dominated by message DEGREE, since it may be broadcasted Δ times by a host, where Δ is the maximum degree. Thus the message complexity of Phase 1 is $O(n \cdot \Delta)$. The time complexity of Phase 1 is $O(n)$.

In phase 2, vertices are explored one by one. The total number of vertices explored is the size of the output CDS. Thus the time complexity is at most $O(n)$. The message complexity is dominated by message NUMOFBLACKNEIGHBORS, which is broadcasted at most 5 times by each gray vertex because a gray vertex has at most 5 black neighbors in a unit-disk graph. Thus the message complexity is also $O(n)$.

From the above analysis, we have

Theorem 3.1. *The distributed algorithm has time complexity $O(n)$ and message complexity $O(n \cdot \Delta)$.*

Note that in phase 1 if we use (id) instead of (d^*, id) as the parameter to select a white vertex to color it black, the message complexity will be $O(n)$ because no DEGREE messages will be broadcasted. $O(n \cdot \Delta)$ is the best result we can achieve if effective degree is taken into consideration.

3.2. Performance Analysis

In this subsection, we study the performance of our algorithm.

Lemma 3.2. *Phase 1 computes an MIS which contains all black nodes.*

Proof. A node is colored black only from white. No two white neighbors can be colored black at the same time since they must have different (d^*, id) .

When a node is colored black, all of its neighbors are colored gray. Once a node is colored gray, it remains in color gray during Phase 1. ■

From the proof of Lemma 3.2, it is clear that if (id) instead of (d^*, id) is used, we still get an MIS. Intuitively, this would yield an MIS of a larger size.

Lemma 3.3. *In phase 2, at least one gray vertex which connects to maximum number of black vertices will be selected.*

Proof. Let u be a gray vertex with maximum number of black neighbors. At some step in phase 2, one of u 's black neighbors v will be explored. In the following step, u will be explored. This exploration is triggered by v . ■

Lemma 3.4. *If there are c black hosts after phase 1, then at most $c - 1$ gray hosts will be colored black in phase 2.*

Proof. In phase 2, the first gray vertex selected will connect to at least 2 black vertices. In the following steps, any newly selected gray vertex will connect to at least one new black vertex. ■

Lemma 3.5. *If there exists a gray vertex which connects to at least 3 black vertices, then the number of gray vertices which are colored black in phase 2 will be at most $c - 2$, where c is the number of black vertices after phase 1.*

Proof. From Lemma 3.3, at least one gray vertex with maximum black neighbors will be colored black in phase 2. Denote this vertex by u . If u is colored black, then all of its black neighbors will choose u as its dominator. Thus, the selection of u causes more than 1 black hosts to be connected. ■

Theorem 3.6. *Our algorithm has performance ratio at most 8.*

Proof. From Lemma 3.2, phase 1 computes an MIS. We will consider two cases here.

If there exists a gray vertex which has at least 3 black neighbors after phase 1, from Lemma 2.1, the size of the MIS is at most $4 \cdot opt + 1$. From lemma 3.5, we know the total number of black vertices after phase 2 is at most $4 \cdot opt + 1 + ((4 \cdot opt + 1) - 2) = 8 \cdot opt$.

If the maximum number of black neighbors a gray vertex has is 2, then the size of the MIS computed in phase 1 is at most $2 \cdot opt$ since any vertex in MCDS connects to at most 2 vertices in the MIS. Thus from Lemma 3.4, the total number of black hosts will be $2 \cdot opt + 2 \cdot opt - 1 < 4 \cdot opt$. ■

Note that from the proof of Theorem 3.6, if (id) instead of (d^*, id) is used in phase 1, our algorithm still has performance ratio at most 8.

We compare the theoretical performance of our algorithm with the algorithms proposed in [1, 8, 17] in Table 3.1. The parameters used for comparison include the (upper bound of the) cardinality of the generated CDS (CDSC), the message and time complexities (MC and TC, respectively), the message length (ML) and neighborhood information (NI).

Table 3.1. Performance comparison of the algorithms in [8, 17, 1] and the one proposed in this chapter. Here opt is the size of the MCDS; Δ is the maximum degree; $|C|$ is the size of the generated connected dominating set; m is the number of edges; n is the number of hosts.

	[8]-I	[8]-II	[17]	[1]	A
CDSC	$\leq (2\ln\Delta + 3)opt$	$\leq (2\ln\Delta + 2)opt$	N/A	$\leq 8opt + 1$	$< 8opt$
MC	$O(n C + m + n\log n)$	$O(n C)$	$O(n\Delta)$	$O(n\log n)$	$O(n)$
TC	$O((n + C)\Delta)$	$O(C (\Delta + C))$	$O(\Delta^2)$	$O(n\Delta)$	$O(n\Delta)$
ML	$O(\Delta)$	$O(\Delta)$	$O(\Delta)$	$O(1)$	$O(1)$
NI	2-hop	2-hop	2-hop	1-hop	1-hop

Note that the last column (labeled by A) in Table 3.1 corresponds to our algorithm. We see our algorithm is superior over the two algorithms in [8] for all parameters. Algorithm in [17] takes less time than our algorithm but it has much higher message complexity and it uses more complicated message information. The algorithm in [1] is comparable with our algorithms in many parameters. But the simulation results in Section 4 show that our algorithm computes smaller in average connected dominating sets for both random and uniform graphs.

4. Simulation

Table 3.1 in the previous section compares our algorithms with others in [1, 8, 17] theoretically. In this section, we will compare the size of the CDSs computed by different algorithms. As mentioned earlier, the virtual backbone is mainly used to disseminate control packets. Thus the most important parameter is the number of hosts in the virtual backbone after it is constructed. The larger the size of a virtual backbone, the larger number of transmissions to broadcast a message to the whole network is needed. Note that the message complexities of the algorithms in [8] and [17] are too high compared to other algorithms and they need 2-hop neighborhood information. Thus we will not consider them in the simulation study. We will compare our algorithm with the one given by [1].

We will consider two kinds of topologies: random and uniform. We assume there are N hosts distributed randomly or uniformly in a 100×100 square

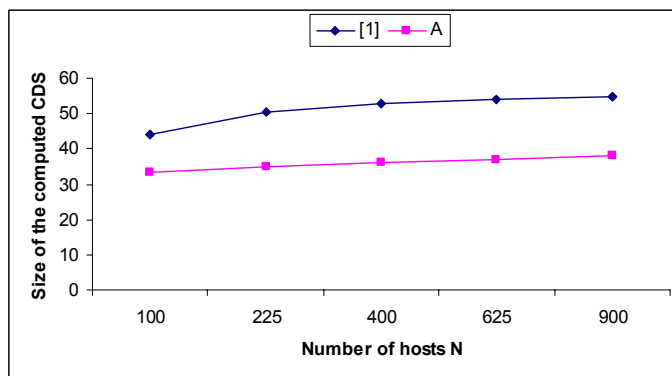


Figure 3.1. Averaged results for $R=15$ in random graphs.

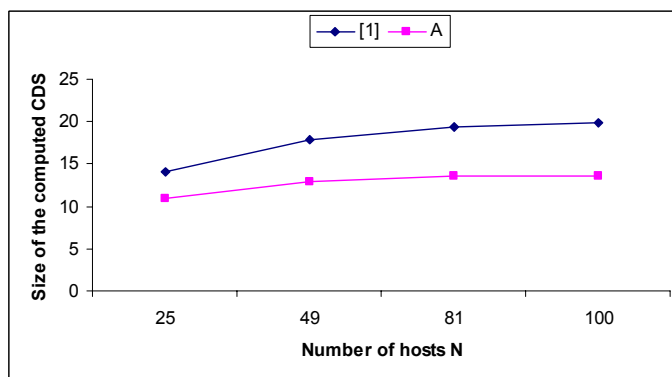


Figure 3.2. Averaged results for $R=25$ in random graphs.

units. Transmission range R is chosen to be 15, 25 or 50 units. For each value of R , we run our algorithms 100 times for different values of N . The averaged results are reported in Figures 3.1, 3.2 and 3.3 for random graphs, and in Figures 3.4, 3.5 and 3.6 for uniform graphs. From these figures it is clear that in all of our simulation scenarios, our algorithm performs better than the algorithm in [1].

5. Conclusion

In this chapter we provide a distributed algorithm which computes a connected dominating set of a small size. Our algorithm has performance ratio at most 8 which is the best to our knowledge. This algorithm takes time $O(n)$ and message $O(n \cdot \Delta)$. Our future work is to investigate the performance of vir-

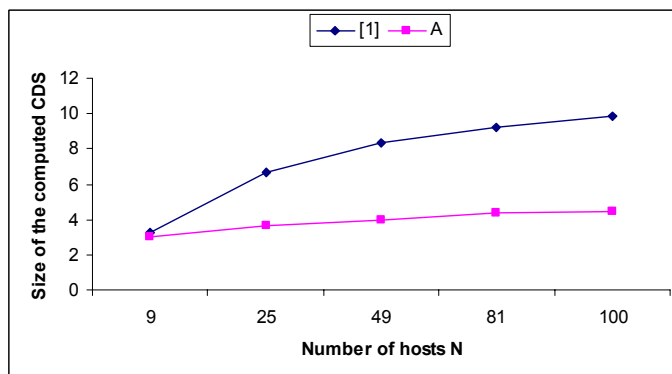


Figure 3.3. Averaged results for $R=50$ in random graphs.

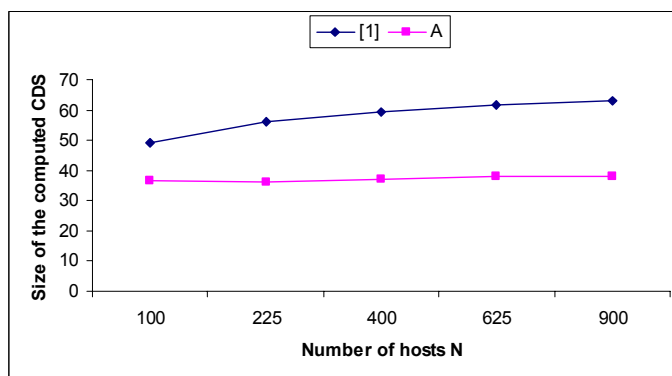


Figure 3.4. Averaged results for $R=15$ in uniform graphs.

tual backbone routing and to study the problem of maintaining the connected dominating set in a mobility environment.

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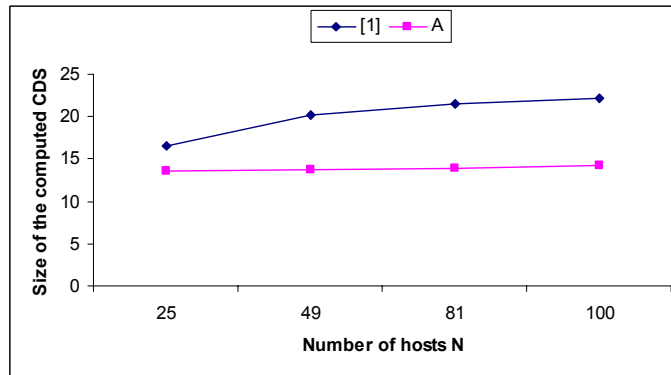


Figure 3.5. Averaged results for $R=25$ in uniform graphs.

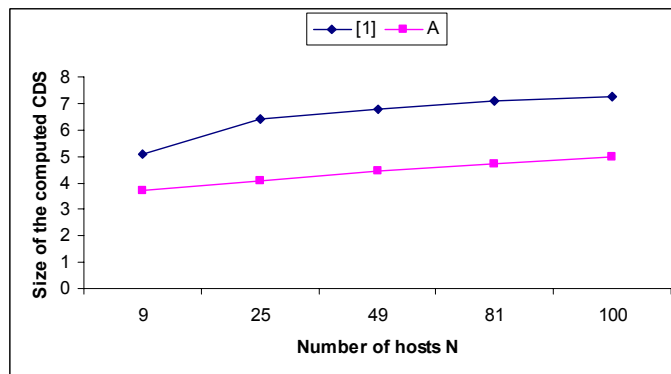


Figure 3.6. Averaged results for $R=50$ in uniform graphs.

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