

# Graph Theoretic Models and Tools for the Analysis of Dynamic Wireless Multihop Networks

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**Abstract**—Wireless multihop networks, in various forms, are being increasingly used in military and civilian applications. Advanced applications of wireless multihop networks demand better understanding on their properties. Existing research on wireless multihop networks has largely focused on static networks, where the network topology is time-invariant; and there is comparatively a lack of understanding on the properties of dynamic networks with dynamically changing topology. In this paper, we use and extend a recently proposed graph theoretic model, i.e. *evolving graphs*, to capture the characteristics of such networks. We extend and develop the concepts of route matrix, connectivity matrix and probabilistic connectivity matrix as a convenient tool to characterize and investigate the properties of evolving graphs and the associated dynamic networks. The properties of these matrices are established and their relevance to the properties of dynamic wireless multihop networks are introduced.

**Index Terms**—Wireless multihop networks, evolving graphs, connectivity matrix

## I. INTRODUCTION

Wireless multihop networks, in various forms e.g. wireless mesh networks, opportunistic networks, delay tolerant networks, mobile ad-hoc networks, vehicular ad-hoc networks and wireless sensor networks, are being increasingly used in military and civilian applications [4]. In multihop networks, wireless devices are self-organized to create a network by exploiting their wireless interfaces and packets are forwarded hop-by-hop by the wireless nodes collaboratively from the source to the destination. Wireless multihop networks can be broadly classified into static networks, where the network topology is time-invariant, and dynamic networks with dynamically changing topology. Advanced applications of wireless multihop networks demand better understanding on the fundamental properties of these networks.

Despite significant research into the properties of static wireless multihop networks, there is much less understanding on the properties of dynamic multihop networks. “Most results found in the literature either deal with networks that are static or are hardly reproducible” [8]. This lack of understanding on

the properties of dynamic multihop networks is attributable to a lack of a set of formal tools to carry out analysis [12].

In wireless multihop networks a change in topology can be caused by node mobility, by node/link failure or by nodes being switched on/off for energy saving purposes. Topology change caused by temporal changes in wireless links is not included because this effect can usually be addressed by a modification of the models and analysis for static wireless multihop networks.. There are two major differences between a static network and a dynamic network:

- In dynamic networks, it is possible that two nodes may never be part of the same connected component but they are still able to communicate with each other. This is illustrated in Fig. 1.
- In dynamic networks, while any one wireless link may be (or may be assumed to be) unidirectional, the path connecting any two nodes must be regarded as directional, i.e. the fact that there is a path from node  $v_i$  to node  $v_j$  within a designated time period does not necessarily mean there is a path from  $v_j$  to  $v_i$  within the same period. For example, in Fig. 1 a message from  $v_6$  can reach  $v_1$  at  $t_2$  but a message from  $v_1$  can only reach  $v_6$  at  $t_4$ .

Therefore many established concepts in static networks must be revisited for dynamic networks. For example, a static wireless multi-hop network is said to be *connected* if there is a path between any pair of nodes in the network. A static wireless multi-hop network is *k-connected* if there are  $k$  mutually independent paths, i.e. paths not having any nodes in common except the starting and ending nodes, between any pair of nodes in the network. However a more meaningful definition of connectivity in dynamic networks is to say that a dynamic network is *connected in time period*  $[0, T]$  if there is a path *from* any node *to* any other node within  $[0, T]$  <sup>1</sup>. The above definition implies that the tradeoff between connectivity, mobility (or active period in networks with stationary nodes but nodes may be switched on/off) and delay must be properly considered in the analysis of dynamic networks. Indeed, it was shown in [9] that in a mobile network with no delay requirement, mobility of nodes can

<sup>1</sup>National ICT Australia is funded by the Australian Government Department of Communications, Information Technology and the Arts and the Australian Research Council through the *Backing Australia Ability initiative* and the ICT Centre of Excellence Program.

<sup>1</sup>A rigorous definition of connectivity in dynamic networks is given in Section III.

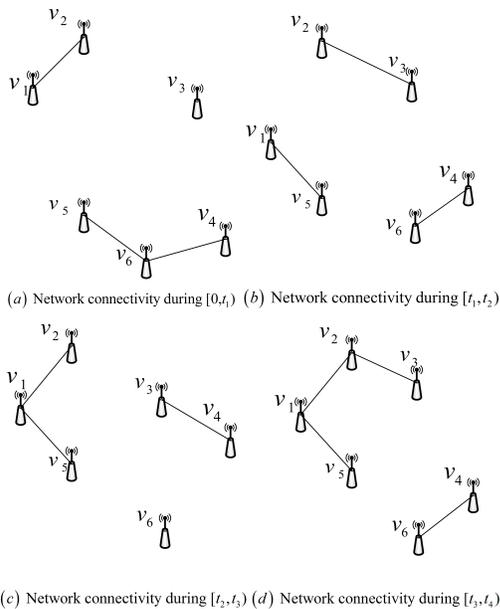


Figure 1. An illustration of connectivity in mobile ad-hoc networks. A solid line represents a direct connection between two nodes. The network is disconnected at any time instant but there is a path from any node to any other node in the network. For example, nodes  $v_1$  and  $v_6$  are never part of the same connected component but a message from  $v_1$  can still reach  $v_6$  through the following path:  $[0, t_1) : v_1 \rightarrow v_2, [t_1, t_2) : v_2 \rightarrow v_3, [t_2, t_3) : v_3 \rightarrow v_4, [t_3, t_4) : v_4 \rightarrow v_6$ .

be exploited to significantly increase network capacity [9], [12]. Similarly it can also be expected that mobility can be exploited to improve connectivity. It remains an open problem to investigate the connectivity and other properties, e.g. end-to-end delay, throughput, of dynamic multihop wireless networks.

In this paper, we use and extend a recently proposed graph theoretic model, i.e. *evolving graphs*, to capture the characteristics, in particular the connectivity, of dynamic wireless multihop networks. We extend and develop a set of tools, i.e. route matrix, connectivity matrix and probabilistic connectivity matrix, to investigate the properties of dynamic multihop wireless networks. The properties of these matrices related to dynamic multihop networks are established. The rest of the paper is organized as follows. In Section II we briefly review related work in the area. In Section III we introduce and extend the concept of evolving graphs as a formal abstraction of dynamic networks. In Section IV we develop route matrix, connectivity matrix and probabilistic connectivity matrix as a convenient algebraic tool to investigate evolving graphs. The properties of these matrices are established. Finally conclusions and further work are given in Section V.

## II. RELATED WORK

A graphic model of a wireless multihop network can be obtained by representing a node in the network uniquely by a vertex in the graph and by representing a wireless link uniquely by an edge in the graph, and vice versa. For simplicity, it is often assumed that the wireless link is symmetric and can be represented by an undirected edge [16],

[12], [15]. The obtained undirected graph  $G(V, E)$  where  $V$  is the vertex set and  $E$  is the edge set is called the *underlying graph* of the wireless multihop network. A widely used model for static wireless multihop networks is *random geometric graphs* (RGG) [16]. In *random geometric graphs*, vertices are *i.i.d.* distributed in a given area following either uniform distribution or Poisson distribution. An edge exists between two vertices iff their Euclidean distance is less than or equal to a given threshold, known as the transmission range. This communication model is known as the *unit disk communication model*.

In the late 1990s and early 2000s, Penrose [16], Gupta and Kumar [11], and Appel and Russo [1] investigated the connectivity of static wireless multihop networks using RGG and showed that in a network consisting of  $n$  nodes identically, independently, randomly and uniformly distributed on a disk of unit area in  $\mathbb{R}^2$ , where any two nodes in the network are directly connected if and only if (iff) their Euclidean distance is smaller than or equal to the transmission range, the critical transmission range  $r_c(n)$  for the network to be connected is  $\sqrt{\frac{\log n}{\pi n}}$  in the sense that the network with  $r(n) = \sqrt{\frac{\log n + c(n)}{\pi n}}$  is asymptotically connected with probability one iff  $c(n) \rightarrow +\infty$  as  $n \rightarrow \infty$ ; the network is asymptotically disconnected with probability one iff  $c(n) \rightarrow -\infty$  as  $n \rightarrow \infty$ . Penrose further demonstrated [16] that in such a network, the critical transmission range  $r_k(n)$  required for the network to be  $k$ -connected and the critical transmission range  $\tau_k(n)$  required for the network to have a minimum node degree  $k$  converge to the same value as  $n \rightarrow \infty$ , i.e.  $\lim_{n \rightarrow \infty} \Pr \{r_k(n) = \tau_k(n)\} = 1$ . These fundamental results on network connectivity form the basis of the work by hundreds of researchers on designing energy efficient routing and scheduling algorithms, topology control and coverage control, localization, and radio resource management, see [13], [20] and references therein for example. There are many other results on the connectivity of static multihop networks. These results [16], [11], [1] on network connectivity apply only to static networks, it remains an open problem to investigate the connectivity of dynamic multihop wireless networks.

The most related work to ours includes the use of *adjacency matrix* to study the connectivity of networks by Cvetkovic et al. [5] and Brooks et al. [3]. In [5], the *adjacency matrix* of a static network of  $n$  nodes is defined to be a  $n \times n$  square matrix  $M$  whose  $(i, j)$ -th entry  $m_{ij} = 1$  if there is a direct connection between distinct nodes  $i$  and  $j$ , otherwise  $m_{ij} = 0$ . Further  $m_{ii}$  is defined to be 0, i.e. no loop is allowed. It is shown that the number of *walks* of length  $z$  between nodes  $i$  and  $j$  is  $m_{ij}^z$ , where  $m_{ij}^z$  is the  $(i, j)$ -th entry of  $M^z$ . A walk of length  $z$  between nodes  $i$  and  $j$  is a sequence of  $z + 1$  vertices  $i, v_1, v_2, \dots, v_{z-1}, j$  which starts at  $i$  and ends at  $j$  and there is a direct connection between adjacent vertices. A *path* of length  $z$  between nodes  $i$  and  $j$  is a walk of length  $z$  between nodes  $i$  and  $j$  in which the vertices  $i, v_1, v_2, \dots, v_{z-1}$  are distinct. If  $i$  is the same as  $j$ , the path is called a *cycle*. In [3] Brooks et al. consider a probabilistic version of the adjacency matrix

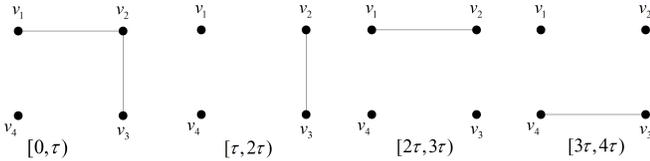


Figure 2. An illustration of the connectivity of a dynamic network in different time intervals. Each subfigure represents the connectivity of a four-node network in a corresponding time interval. A solid line represents a direct connection between two nodes.

for multihop networks and define a *probabilistic adjacency matrix*<sup>2</sup> as a  $n \times n$  square matrix  $M$  whose  $(i, j)$ -th entry  $m_{ij}$  represents the probability of having a direct connection between distinct nodes  $i$  and  $j$ , and  $m_{ii} = 0$ . They established that the probability that there exists at least one walk of length  $z$  between nodes  $i$  and  $j$  is  $m_{ij}^z$ , where  $m_{ij}^z$  is the  $(i, j)$ -th entry of  $M \otimes M \otimes \dots \otimes M$  ( $z$  times).  $C \doteq A \otimes B$  iff  $C_{ij} = 1 - \prod_{l \neq i, j} (1 - A_{il}B_{lj})$  where  $A_{ij}$ ,  $B_{ij}$  and  $C_{ij}$  are the  $(i, j)$ -th entries of the  $n \times n$  square matrix  $A$ ,  $B$  and  $C$  respectively.

The above adjacency matrix and probabilistic adjacency matrix, however, is not suitable for dynamic networks. This is illustrated through the example dynamic network of four nodes shown in Fig. 2. Denote the adjacency matrix of the network during  $[(i-1)\tau, i\tau)$  by  $M_i$ . Following the above definition of adjacency matrix, it can be concluded that there is no connection from  $v_1$  to  $v_4$  in the network within  $[0, 4\tau)$  because the  $(1, 4)$ -th entries in  $M_1$ ,  $M_1M_2$ ,  $M_1M_2M_3$  and  $M_1M_2M_3M_4$  are all zero. This is obviously incorrect because a message from  $v_1$  can reach  $v_4$  using the following journey<sup>3</sup>:  $[0, \tau) : v_1 \rightarrow v_3$ ,  $[3\tau, 4\tau) : v_3 \rightarrow v_4$ ; or if we further impose the limit that a message can only travel one hop during each time interval, a message from  $v_1$  can reach  $v_4$  using the journey  $[0, \tau) : v_1 \rightarrow v_3$ ,  $[\tau, 3\tau) : v_2 \rightarrow v_3$ ,  $[3\tau, 4\tau) : v_3 \rightarrow v_4$ . If the 1s in  $M_i$ ,  $1 \leq i \leq 4$  are replaced by values between 0 and 1 representing the probabilities that there are direct connections between the corresponding pairs of vertices,  $M_i$ ,  $1 \leq i \leq 4$  becomes the probabilistic adjacency matrix of the network during  $[(i-1)\tau, i\tau)$  as defined in [3]. It can be obtained that the probability that there exists a connection from  $v_1$  to  $v_4$  in the network within  $[0, 4\tau)$  is zero because the  $(1, 4)$ -th entries in  $M_1$ ,  $M_1 \otimes M_2$ ,  $M_1 \otimes M_2 \otimes M_3$  and  $M_1 \otimes M_2 \otimes M_3 \otimes M_4$  are all zero. Apparently this is not true. Therefore neither the adjacency matrix defined in [5] and the probabilistic adjacency matrix defined in [3] is suitable for investigating the connectivity of dynamic networks.

Other related work includes [17] and [6]. In [17] Santi and Blough investigated through simulations the connectivity of mobile ad-hoc networks for different combinations of number of nodes  $n$  and transmission range  $r$  using the random walk mobility model [10]. In [6] Díaz et al. analyzed the connectivity of a mobile ad-hoc network in which a total of

<sup>2</sup>It is called connectivity matrix in [3] but is more widely referred to as adjacency matrix in the literature. In this paper, we call it adjacency matrix to differentiate it from the connectivity matrix defined later in the paper.

<sup>3</sup>We will give a rigorous definition of the term *journey* in Section III.

$m$  nodes move randomly along the edges of a  $n \times n$  grid.

### III. EVOLVING GRAPH THEORY

A major challenge in the analysis of dynamic networks is choosing a proper graphic model that can both capture the dynamics of the network and be tractable for analytical studies. We use and extend a class of graphs, *evolving graphs* [2], [8], [19], as a formal abstraction of dynamic networks for our analysis. There has been research in the literature on the evaluation of computational complexity of algorithms in evolving graphs [19], [2] and on simulation based studies on the performance of routing algorithms in dynamic networks using evolving graphs [14].

An *evolving graph*  $\mathcal{G} = (G, S_G)$  consists of a given graph  $G(V, E)$ , along with an ordered sequence of its subgraphs,  $S_G = G_1(V_1, E_1), G_2(V_2, E_2), \dots, G_\tau(V_\tau, E_\tau)$ ,  $\tau \in \mathbb{N}$ . Let the vertex set and the edge set of  $\mathcal{G}$  be  $V_G = \cup V_i$  and  $E_G = \cup E_i$  respectively. If we assign a physical meaning to subgraphs such that the subgraph  $G_i(V_i, E_i)$  at a given index  $i$  is the *underlying graph* of the network during the time interval  $[t_{i-1}, t_i)$  and  $t_0 < t_1 < \dots < t_\tau$ , the time domain is then incorporated into the model. Let  $P_i$  be a path in  $G_i$ . Let  $F(P_i)$  and  $L(P_i)$  be the first vertex and the last vertex of  $P_i$  respectively and  $|P_i|$  be its length. We define a *journey* in  $\mathcal{G} = (G, S_G)$  from vertex  $u$  to vertex  $v$  of  $V_G$  as a sequence  $J(u, v) = P_{\lambda_1}, P_{\lambda_2}, \dots, P_{\lambda_l}$ , with  $\lambda_1 < \lambda_2 < \dots < \lambda_l \leq \tau$ , such that is  $P_{\lambda_i}$  a path in  $G_{\lambda_i}$  with  $F(P_{\lambda_1}) = u$ ,  $L(P_{\lambda_l}) = v$ , for all  $1 \leq i \leq l$ , it holds that  $F(P_{\lambda_i}) \neq L(P_{\lambda_i})$  and all  $i < l$ , it holds that  $L(P_{\lambda_i}) = F(P_{\lambda_{i+1}})$ . The time domain is further incorporated into the model by restricting journeys to never move into paths (or edges) that existed only in past subgraphs. The length of the journey is  $\sum_{i=1}^l |P_{\lambda_i}|$ . The restriction  $F(P_{\lambda_i}) \neq L(P_{\lambda_i})$  has limited the journey to consist of non-cycle paths only. Nevertheless a vertex may still appear more than once inside a journey, viz a vertex might appear both in a path  $P_{\lambda_i} \in J(u, v)$  and in a path  $P_{\lambda_j} \in J(u, v)$ , where  $\lambda_i \neq \lambda_j$ . The starting and the ending vertices of a journey may also be the same. We define an *evolving path* in  $\mathcal{G} = (G, S_G)$  from vertex  $u$  to vertex  $v$  of  $V_G$  as a journey  $J(u, v)$  in which all vertices are distinct except that the ending vertex of the journey may be the same as the starting vertex. The term *evolving path* is intended to differentiate a path in an evolving graph from a path in one of the subgraphs  $G_{\lambda_i}$ . A vertex  $v$  is said to be  $k$ -hops away from another vertex  $u$  if the length of the shortest journey (or equivalently the shortest evolving path) from vertex  $u$  to vertex  $v$  is  $k$ . The evolving graph  $\mathcal{G} = (G, S_G)$  is said to be a *connected* graph if there exists a journey (or equivalently evolving path) from any vertex to any other vertex of  $V_G$ .

### IV. CONNECTIVITY MATRIX AND PROBABILISTIC CONNECTIVITY MATRIX FOR DYNAMIC NETWORKS

To facilitate later discussions, we call a dynamic network *deterministic* if all entries of its adjacency matrix, as defined in [5], during each time interval are either 1 or 0, representing there are or there are no direct connections between

the corresponding pairs of vertices respectively. We call a dynamic network *probabilistic* if during some time intervals only statistical information on the probabilities that there are direct connections between certain pairs of nodes is known.

#### A. Connectivity Matrix of Deterministic Dynamic Networks

Let  $0, \tau_1, \tau_2, \dots, \tau_k, \dots, 0 < \tau_1 < \tau_2 < \dots < \tau_k \dots$  be the embedding points of the network such that network topology, defined by a collection of all direct connections between nodes in the network, only changes at these discrete time instants. We define the *route matrix*  $H_k$  of a deterministic dynamic network of  $n$  nodes during time interval  $[\tau_{k-1}, \tau_k)$  whose underlying graph is  $G_k$  as a  $n \times n$  square matrix such that its  $(i, j)$ -th entry is the number of paths in  $G_k$  from vertex  $i$  to another *distinct* vertex  $j$ . If there is no path from  $i$  to  $j$ ,  $(H_k)_{ij} = 0$  where  $(H_k)_{ij}$  is the  $(i, j)$ -th entry of  $H_k$ . Further the diagonal entries of  $H_k$  is 1 by *definition*. This definition of the diagonal entries reflects the fact that in dynamic networks routes and connections are constantly changing, when a node either becomes isolated or cannot find a (better) route to the destination, it has to store the message until a (better) route is available. The implication of such definition will become clear in later discussions. For convenience, we assume that  $V_1 = V_2 = \dots = V_k = \dots$  such that the route matrices during different time intervals have the same dimension. If a node disappears and then reappears in the network, the situation can be readily accommodated by considering that node as an isolated node during the corresponding time interval.

Based on the above definition of the route matrix, we can establish some properties of the route matrix and their relevance to the properties of the dynamic network.

*Lemma 1:* There are a total of  $H_{ij}^k$  journeys from node  $i$  to node  $j$  in the evolving graph  $\mathcal{G} = (G, S_G)$ ,  $S_G = G_1(V_1, E_1), G_2(V_2, E_2), \dots, G_k(V_k, E_k)$ ,  $k \in \mathbb{N}$ , where  $H_{ij}^k$  is the  $(i, j)$ -th entry of  $H_1 H_2 \dots H_k - I$ .

*Proof:* We prove Lemma 1 by iteration.

When  $k = 1$ , the lemma is obviously correct. Note that the diagonal entries of  $H_1 - I$  is 0, i.e. there is no journey from a node to itself within  $[0, \tau_1)$ .

When  $k = 2$ ,

$$H_{ii}^2 = h_{ii}^1 h_{ii}^2 + \sum_{l \neq i} h_{il}^1 h_{li}^2 - 1 = \sum_{l \neq i} h_{il}^1 h_{li}^2$$

is the total number of journeys from node  $i$  to itself within  $[0, \tau_1) \cup [\tau_1, \tau_2)$  (each journey is made of two non-cycle path, one from  $i$  to  $l \neq i$  in  $G_1$  and the other from  $l$  to  $i$  in  $G_2$ ), where  $h_{ij}^p$  is the  $(i, j)$ -th entry of  $H_p$ . For distinct  $i$  and  $j$ ,

$$H_{ij}^2 = \sum_l h_{il}^1 h_{lj}^2 = h_{ij}^1 + h_{ij}^2 + \sum_{l \neq i, j} h_{il}^1 h_{lj}^2.$$

The first term in the above equation is the number of journeys from node  $i$  to node  $j$  made of paths in  $G_1$  only; the second term is the number of journeys from  $i$  to  $j$  made of paths in  $G_2$  only; and the third term is the number of journeys from node  $i$  to node  $j$ , where each journey is made of a non-cycle

path in  $G_1$  and a non-cycle path in  $G_2$ . Therefore the lemma is true for  $k = 2$ .

Assume the lemma is correct for  $k = T$ , let us consider  $k = T + 1$  and  $i \neq j$ .

$$H_{ij}^{T+1} = \sum_l H_{il}^T h_{lj}^{T+1} = H_{ij}^T + H_{ii}^T h_{ij}^{T+1} + \sum_{l \neq i, j} H_{il}^T h_{lj}^{T+1}.$$

In the above equation, the first term is the number of journeys from  $i$  to  $j$  within  $[0, \tau_T)$ . The sum of the second and the third terms is the number of newly added journeys from  $i$  to  $j$  where each journey consists of a non-cycle path in  $G_{T+1}$ . The second term is the sum of  $(H_{ii}^T - 1) h_{ij}^{T+1}$ , representing the number of journeys that start at  $i$  before  $\tau_T$ , end at  $i$  at  $\tau_T$ , and reach  $j$  at  $\tau_{T+1}$ , and  $h_{ij}^{T+1}$ , representing the number of journeys from  $i$  to  $j$  made of non-cycle paths in  $G_{T+1}$  only; and the third term is the number of journeys that end at an intermediate node at  $\tau_T$  and reach  $j$  at  $\tau_{T+1}$ . The sum of the three terms gives the total number of journeys from node  $i$  to node  $j$  within  $[0, \tau_{T+1})$ . It is trivial to show that  $H_{ii}^{T+1} - 1$  is the total number of journeys that start and end at  $i$  within  $[0, \tau_{T+1})$ . ■

*Remark 1:* Denote by  $H'_k$  the route matrix of  $G_k$  except that the diagonal entries of  $H'_k$  are zeros, i.e.  $H'_k = H_k - I$ . It can be shown that

$$H_1 H_2 \dots H_k - I = \sum_i H'_i + \sum_{i < j} H'_i H'_j + \dots + H'_1 H'_2 \dots H'_k.$$

The  $i$ -th term on the right-hand side of the above equation represents the number of journeys with exactly  $i$  non-cycle paths in  $i$  distinct subgraphs of  $\mathcal{G}$  respectively. Therefore the above definition of the route matrix and Lemma 1 allows the preservation of the history of journeys ever occurred in the evolving graph  $\mathcal{G}$ .

*Lemma 2:* A dynamic network is connected in  $[0, \tau_k)$  iff  $H_1 H_2 \dots H_k > J$  element wise, where  $J$  is a  $n \times n$  matrix with all entries equal to 1.

The proof of Lemma 2 is omitted.

*Lemma 3:*  $H_{ij}^k$  is a non-decreasing function of  $k$ .

This lemma is an easy consequence of Lemma 1. An implication of Lemma 3 is that in dynamic networks the number of journeys from any node to any other node will only increase over time.

We define the *connectivity matrix*  $Q_k$  of a deterministic dynamic network of  $n$  nodes whose underlying graph is  $G_k$  during time interval  $[\tau_{k-1}, \tau_k)$  as a  $n \times n$  square matrix such that its  $(i, j)$ -th entry is 1 if there exists a path in  $G_k$  from vertex  $i$  to another distinct vertex  $j$ . If there is no path from  $i$  to  $j$ ,  $(Q_k)_{ij} = 0$ .  $(Q_k)_{ii} = 1$ . Using the properties of non-negative matrices [18] and the above lemmas, we can establish the following properties of the connectivity matrix.

*Lemma 4:* A dynamic network is connected in  $[0, \tau_k)$  iff  $Q_1 Q_2 \dots Q_k \geq J$  element wise. There is a journey from vertex  $i$  to  $j$  iff  $(Q_1 Q_2 \dots Q_k)_{ij} \geq 1$ .

It can also be established that  $(Q_1 Q_2 \dots Q_k)_{ij}$  is a non-decreasing function of  $k$ . This, together with Lemma 4, suggests for a mobile network in which mobile nodes travel

according to a random mobility model, e.g. random walk model [10], the network will become connected eventually. The interesting question is thus the rate at which the network becomes connected or equivalently the rate at which all entries of  $Q_1 Q_2 \cdots Q_k$  (or equivalently  $H_1 H_2 \cdots H_k$ ) become greater than or equal to 1.

The connectivity matrix and the route matrix defined earlier, although are a convenient tool to investigate connectivity, do not capture the fact that in a communications network a message can only travel a limited distance within a limited amount of time. In view of this, we extend the earlier definitions to account for message propagation delays. We call a path  $P_i$  in  $G_i$  *feasible* if a message of a unit length transmitted by the starting point of  $P_i$  at  $\tau_{i-1}$  can reach the end point of  $P_i$  before  $\tau_i$ . We call a journey *feasible* if it is made entirely of feasible paths. We can also define the concept of *feasible connectivity matrix* and *feasible route matrix* analogously. Obviously the above lemmas on the properties of connectivity matrix and route matrix still hold if the relevant terms are replaced by their “feasible” counterparts. By a direct application of Lemma 1, we obtain the following Lemma that relates the feasible connectivity matrix to end-to-end throughput (latency).

*Lemma 5:* Assuming that node  $i$  starts to transmit a message of unit length to another distinct node  $j$  at time 0, there are  $M_{ij}^k$  feasible journeys from  $i$  to  $j$  that take less than or equal to  $\tau_k$  time for the unit message to reach  $j$ , where  $M_{ij}^k$  is the  $(i, j)$ -th entry of  $M_1 M_2 \cdots M_k$ .  $M_k$  is the feasible route matrix of graph  $G_k$  during  $[\tau_{k-1}, \tau_k)$ .

Lemma 5 can also be used to derive an upper bound on the end-to-end throughput in a dynamic network. Further let  $\tau$  be the time required to transmit a message of unit length from a node to its immediate neighbors. We choose the embedding points of the network  $0, \tau_1, \tau_2, \dots, \tau_k, \dots$  to be  $\tau_k = k\tau$  and assume that a node can only start transmitting a message at  $\tau_k$  which accords with a synchronized network. If a direct connection between two nodes is broken during the interval  $[(i-1)\tau, i\tau)$ , it is considered that there is no direct connection between the two nodes during  $[(i-1)\tau, i\tau)$ . Under the above conditions, a *feasible connectivity matrix* becomes the sum of the corresponding adjacency matrix and  $I$ . We can obtain the following lemma:

*Lemma 6:* Assuming that node  $i$  starts to transmit a unit message to another distinct node  $j$  at time 0, the least amount of time for the message to arrive at node  $j$  is  $k\tau$  iff  $M_{ij}^{k-1} = 0$  and  $M_{ij}^k \geq 1$ . The maximum number of messages that can be transmitted from  $i$  to  $j$  within  $[0, k\tau)$  is  $M_{ij}^k$ .

Lemmas 5 and 6 give an easy way to study the *foremost* journeys [19], [2] in an evolving graph. A slight modification of the two lemmas can be used to investigate the *shortest* journeys and the *fastest* journeys [19], [2] in an evolving graph algebraically.

### B. Probabilistic Connectivity Matrix of Probabilistic Dynamic Networks

Let  $0, \tau_1, \tau_2, \dots, \tau_k, \dots, 0 < \tau_1 < \tau_2 < \dots < \tau_k \cdots$  be the embedding points of the network such that network topology

only changes at these discrete time points. We define the *probabilistic connectivity matrix*  $Q_k$  of a probabilistic dynamic network of  $n$  nodes during time interval  $[\tau_{k-1}, \tau_k)$ , whose underlying graph is  $G_k$ , as a  $n \times n$  square matrix such that its  $(i, j)$ -th entry is the probability that there is a path in  $G_k$  from vertex  $i$  to another distinct vertex  $j$  and  $(Q_k)_{ii} \doteq 1$ . We further assume that a) the events represented by entries in the same probabilistic connectivity matrix are independent; b) the events represented by entries in probabilistic connectivity matrices during different time intervals are independent. An example in which these assumptions hold is wireless sensor networks with uncoordinated power saving mechanisms [7]. We can establish the following properties of probabilistic connectivity matrix.

*Lemma 7:* The probability that there exists at least one journey from node  $i$  to another distinct node  $j$  in the evolving graph  $\mathcal{G} = (G, S_G)$ ,  $S_G = G_1(V_1, E_1), G_2(V_2, E_2), \dots, G_k(V_k, E_k)$ ,  $k \in \mathbb{N}$  is  $Q_{ij}^k$ , where  $Q_{ij}^k$  is the  $(i, j)$ -th entry of  $Q_1 \otimes Q_2 \cdots \otimes Q_k$ .  $C = A \otimes B$  means

$$C_{ij} = 1 - \prod_l (1 - A_{il} B_{lj}),$$

and  $A_{ij}$ ,  $B_{ij}$  and  $C_{ij}$  are the  $(i, j)$ -the entries of  $n \times n$  square matrices  $A$ ,  $B$  and  $C$  respectively.

*Proof:* We prove this lemma by iteration. The lemma is obviously correct for  $k = 1$  and the diagonal elements  $Q_{ii}^1 = 1$ ,  $1 \leq i \leq n$ .

Assuming the lemma is correct for  $k = T$ , let us consider  $k = T + 1$ . First, it can be shown that the diagonal elements of  $Q_1 \otimes Q_2 \cdots \otimes Q_{T+1}$  are always 1. Let us now consider  $Q_{ij}^{T+1}$  where  $i \neq j$ ,

$$Q_{ij}^{T+1} = 1 - (1 - q_{ij}^{T+1}) (1 - Q_{ij}^T) \prod_{l \neq i, j} (1 - Q_{il}^T q_{lj}^{T+1})$$

In the above equation,  $q_{ij}^{T+1}$  is the  $(i, j)$ -th entry of  $Q_{T+1}$  and represents the probability of having a journey consisting of a path from  $i$  to  $j$  in  $G_{T+1}$  only.  $Q_{ij}^T$  is the probability of having a journey from  $i$  to  $j$  in the evolving graph  $\mathcal{G} = (G, S_G)$ ,  $S_G = G_1, G_2, \dots, G_T$ . The existence of such a journey necessarily means the existence of a journey in  $\mathcal{G} = (G, S_G)$ ,  $S_G = G_1, G_2, \dots, G_{T+1}$ . The term  $Q_{il}^T q_{lj}^{T+1}$ ,  $l \neq i, j$  is the probability of having a journey in  $\mathcal{G} = (G, S_G)$ ,  $S_G = G_1, G_2, \dots, G_{T+1}$  that ends at an intermediate node  $l$  at  $\tau_T$  and reaches  $j$  at  $\tau_{T+1}$ . Using the assumption on the independence of events represented by entries in the probabilistic connectivity matrix, we can obtain that the above three events are independent. Therefore  $\prod_l (1 - Q_{il}^T q_{lj}^{T+1})$  is the probability that none of the above three events occurs. Finally  $1 - \prod_l (1 - Q_{il}^T q_{lj}^{T+1})$  gives the probability that at least one of the above three events occurs, which is the probability of having a journey from  $i$  to  $j$  in the evolving graph  $\mathcal{G} = (G, S_G)$ ,  $S_G = G_1, G_2, \dots, G_{T+1}$ . ■

We can establish the following properties of the probabilistic connectivity matrix:

*Lemma 8:*  $Q_{ij}^k$  is a non-decreasing function of  $k$ . This lemma can be proved using the definition of  $Q_{ij}^k$ .

*Lemma 9:* A network of  $n$  nodes is connected with probability 1 in  $[0, \tau_k)$  iff  $M_{ij}^k = 1$  for all pairs of  $i$  and  $j$ . This lemma is an easy consequence of Lemma 7.

Similarly by defining the concept of *feasible probabilistic connectivity matrix*, we can use the probabilistic connectivity matrix to investigate end-to-end delay. We define the *feasible probabilistic connectivity matrix*  $Q_k$  of a probabilistic dynamic network of  $n$  nodes during time interval  $[\tau_{k-1}, \tau_k)$  as a  $n \times n$  square matrix such that its  $(i, j)$ -th entry is the probability that there is a *feasible* path in  $G_k$  from vertex  $i$  to another distinct vertex  $j$  and  $(Q_k)_{ij} \doteq 1$ . The above lemmas on the properties of probabilistic connectivity matrix still hold if the relevant terms are replaced by their “*feasible*” counterparts.

Further let  $\tau$  be the time required to transmit a message of unit length from a node to its immediate neighbors. We choose the embedding points of the network  $0, \tau_1, \tau_2, \dots, \tau_k, \dots$  to be  $\tau_k = k\tau$  and assume a node can only start transmitting a message at  $\tau_k$ . If a direct connection is broken during the interval  $[(i-1)\tau, i\tau)$ , it is considered that there is no connection between the two nodes during  $[(i-1)\tau, i\tau)$ . Under the above conditions, a *feasible probabilistic connectivity matrix* becomes the sum of the corresponding probabilistic adjacency matrix, as defined in [3], and  $I$ . We can establish the following lemma:

*Lemma 10:* Assuming that node  $i$  starts to transmit a unit message to a distinct node  $j$  at time 0, the probability that the message can reach  $j$  in less than  $\tau_k$  time is  $Q_{ij}^k$ , where  $Q_{ij}^k$  is the  $(i, j)$ -th entry of  $Q_1 \otimes Q_2 \cdots \otimes Q_k$  and the mean end-to-end delay from  $i$  to  $j$  is given by:

$$\sum_{k=1}^{\infty} (1 - Q_{ij}^{k-1}) Q_{ij}^k k\tau.$$

This lemma can be obtained from Lemma 7. It can be used to study the mean end-to-end delay and the delay distribution.

## V. CONCLUSION AND FURTHER WORK

Wireless multihop networks is an fascinating area that has attracted growing interest. In this paper, we used and extended a recently proposed graph theoretic model, i.e. *evolving graphs*, as a formal abstraction of dynamic wireless multihop networks. We extended and developed a set of tools, i.e. route matrix, connectivity matrix and probabilistic connectivity matrix, to investigate the properties of dynamic multihop wireless networks. The properties of these matrices were investigated in relation to the properties of the corresponding dynamic multihop networks. We expect them to provide a convenient tool to formally investigate the properties of dynamic multihop networks.

An implicit hypothesis used in the paper is that a wireless node is able to store a message when it becomes isolated or cannot find better route to the destination. This has been the case in most wireless multihop networks, especially in an age when memory unit becomes increasing cheap and small-sized. However the tools developed in this paper can be easily modified to account for the situation that a node loses its message because of, e.g. limited battery power or buffer size.

Future work in the area will be carried out along two directions. First, we shall apply the model and tools to investigate the properties of dynamic networks, in particular the asymptotic properties of dynamic networks either when the number of nodes is large or when the time span of the associated evolving graph is long. Second, we shall further develop the tools to make them suitable for a wider category of dynamic multihop networks. For examples, the two independence assumptions in Section IV-B is a bit restrictive. In a mobile multi-hop network, the probabilistic connectivity matrices (adjacency matrices) during adjacent time intervals can be highly correlated. We expect to develop the tools to properly consider the time correlation among matrices.

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