

# Utilizing Colored Pheromones and Helping Ants for Wireless Mesh Networks Routing

Ehsan Ghasemkhani<sup>1</sup>, Rahim Alizadeh<sup>2</sup>, Alireza Mirzaei Nejad Kousari<sup>1</sup>

<sup>1</sup>Department of Computer Science, Islamic Azad University, East Tehran (Ghiamdasht) Branch, Tehran, Iran <sup>2</sup>Department of Mathematics, Shahed University, Tehran, Iran Email: {eh.ghasemkhani, alizadeh.ra}@gmail.com, ali\_mnkt@yahoo.com

Received December 16, 2011; revised January 17, 2012; accepted January 30, 2012

## ABSTRACT

Improving routing algorithm performance not only leads to appreciate the quality of data transmission, but also increases the speed of data transfer. In this paper we propose a hybrid method which is a combination of traffic classification by the help of colored pheromones and *helping ants* method in the intermediate nodes. This combination increases the convergence speed and decreases the delay and Jitter in the network. Also we compare the obtained results with two known routing algorithms that are based on the ant colony.

Keywords: Wireless Mesh Networks; AntNet; Mobile Agents; Network Routing; Helping Ant and Colored Pheromones

## 1. Introduction

A safe routing in the network will lead to appropriate transfer of data from the source to the destination. This has got a high impact on performance of the network in terms of lower delay and jitter as well as higher throughput. In WMNs (Wireless Mesh Networks), nodes are comprised of mesh routers and mesh clients. Each node operates not only as a host but also as a router, forwarding packets on behalf of other nodes that may not be within direct wireless transmission range of their destinations. For more information about routing in WMNs refer to [1-19].

After a generation of the routing algorithms that used to search the shortest route in order to reach convergence, and in them there was a need for updates in the routing tables, a new generation of algorithms were introduced that operated based on multi agent systems.

In the computer network routing applications, most of the agent-based algorithms have been inspired by the life of the ants because of the distributed environment information. By the help of chemical called pheromones, real ants find the shortest path between the food source and their nest. Schonderwoerd *et al.* [20] used agents that are similar to ants. In this algorithm each node sends an ant toward random destination *d* at regular intervals. When the ant reaches node *i*, the next node selects the next hop *j* through routing table for node *i* and after updating the routing table of the node *i* it will go to node *j*. In order to updating the routing table, the probability for selecting *j*  as the next node for destination d should be increased and the probability for selecting other nodes neighboring i should be decreased. Caro and Dorigo [21,22] introduced AntNet algorithm based on the ants routing method. In this algorithm, there are two ant-like agents called forward and backward ants and Like Schonderwoerd's algorithm, each node sends a forward ant toward a specified destination at regular intervals. But unlike the previous algorithm, this ant is only assigned to move toward the concerned destination and obtain information about the path. When arriving at its destination, the forward ant generates another ant, called backward ant and transfers its data to this ant, before dying. The backward ant travels back from the same path as taken by the forward ant and updates the routing table when it gets to each node (with information about the whole path, the backward ant updates the table of nodes, so making better changes in the tables). Camara and Loureiro [23] recommended a routing algorithm called GPSAL (GPS/ ant-like routing algorithm) which is based on GPS (Global Positioning System) and mobile software agents that modeled on ants for routing in ad hoc networks. Soltani et al. [24] added helping ants to the standard AntNet algorithm in the intermediate nodes to update the routing tables, which enhanced the speed of convergence.

But the problem with ants colony based algorithms is that they only use one color (or dimension) of pheromone. Colored Ant System (CAS) [25] introduces "multi-colored pheromones" for solving the issue of graph coloring and distributes agents to the node clusters. The multiple ant colony optimization (MACO) [26] uses multiple ant colonies, each trying to alleviate the inactivity problems and increase the algorithm consistency with the use of specific pheromone colors, but this method does not use the traffic classes. The ARAMA [10] algorithm also tries to justify different aspects of "goodness", but certainly the needs are not classified in it and different sections are not distributed separately and only one pheromone is used. Labella and Dressler [11] applied the colored pheromones to distribute the work among the nodes in the sensor/actuator networks (SANET). Also Umlauft and Elmenreich [27] introduced an algorithm called CPANT, in which the traffic is distributed into different classes by using colored pheromones.

In this present paper we introduce a hybrid algorithm based on the AntNet algorithm that increases not only speed of convergence, but also the quality of the bandwidth service in the network. We suggest the use of two features:

1) Use of a third agent (updating ants of the neighboring nodes) that increases the cooperation among the nodes and reduces delay rate in data transmission.

In order to reach this goal, besides the forward and backward ants, we use updating ants. When a backward ant arrives to a node, and finds path is a good one, it generates updating ants and sends them to one of the neighbors in order to make them informed about the path. Upon arriving at the destination, the updating ant modifies the node and dies (they are not propagated). Adding this type of new ant helps the neighboring nodes to utilize the information and experiences of their neighbors, as nodes have more information to make decisions [24].

2) Use of colored pheromone for classification of traffic. Using this feature, instead of considering a single goodness criterion, several criteria are considered and the data is transmitted according to the color of the pheromone in the concerned path. In this method, according to path features include bandwidth, delay and jitter, a certain path is selected and ants mark paths throughout the network by depositing pheromones with different colors depending on the suitability of the path for the corresponding traffic classes [27].

The structure of the reminder of this Paper is as follows: first in Section 2 we describe standard AntNet algorithm. Our proposed method is introduced in Section 3. Section 4 explains the simulation results and compares them with the standard AntNet algorithm and CPANT algorithm. Finally, in Section 5 we highlight the conclusions of the paper.

## 2. AntNet Algorithm

Ants colony algorithm is based on colony of ants in nature. Ants are able to find their way between their nest and the food source by propagation of pheromone. By propagating pheromone on the path, ants actually inform other ants. Suppose ants get to a point where they should make decision to turn right or left and since they don't know which route is the shortest, they turn left or right by chance (let's assume 50% choose to turn left and 50% choose to turn right). The shortest path is taken by a larger number of ants because the propagated pheromone is more on that path. After a short period of time, the difference of the volume of pheromone existing on the path will increase so that in the next turns, ants choose the route with larger volume of pheromone and higher probability.

By using mobile agents, AntNet in a way simulates the ants work. The pheromone chemical in AntNet algorithm has been simulated with the probabilities existing in the routing tables. When an ant passes through a path, with the use of specified formulas, the possibility of taking the mentioned path will increase and the possibility of taking other routes will decrease.

In this algorithm, ants are divided into two groups of forward and backward ants. The backward ants utilize the information collected by forward ants during their travel from the source to the destination, therefore no node is updated by forward ants, but they report the conditions of network delay to backward ants. This information is the travel time taken between each node of the network. Backward ants use this information in order to update the node routing tables. **Table 1** shows a type of routing table. The items in this routing table are the probabilities and the sum of each row will be equal to one. Ants use these figures to make decision for selecting the next hop [26].

Each node k has got two data structures, which can be updated by agents (N is the total number of the nodes in the network and  $N_k$  is the number of neighbors of node k).

The routing table  $P_k$  has N rows and  $N_k$  columns. In this table, for each pair of nodes (d, n) the probability  $P_{d,n}$  is kept, which is the probability of the goodness of the node n selected as the next node for destination d. Clearly we have

$$\sum_{n=1}^{N_k} p_{d,n} = 1, \ d \in \{1, \cdots, N\}.$$

The vector  $M_k = \left[ \left( \mu_j, \sigma_j^2, T_{best_j} \right) \right], j = 1, \dots, N$ , keeps

#### Table 1. An example of routing table, used by ants.

Destinations –	Neighbor nodes	
Destinations —	В	С
Е	0.15	0.85
F	0.75	0.25

 $\mu_j$ ,  $\sigma_j^2$ ,  $T_{best_j}$  for node k per each node *j*, where they are average, variance and the shortest Time spent by agents in order to travel from *k* to *j*, respectively.

Different Stages of AntNet Algorithm is as follows:

1) Each node of the network sends forward ants to the selected destinations at regular intervals. Based on the local traffic load, the probability of sending forward ant from node s to destination d is obtained with the following formula

$$P_{s,d} = \frac{f_{s,d}}{\sum_{d'=1}^{N} f_{s,d'}}$$

where  $f_{s,d}$  is the volume of the data (the number of bits or packages) that is transferred from node *s* to *d*.

2) The forward ants choose a route based on the current routing tables. Each neighbor is selected as the next node with the probability of  $P'_{n,d}$ , where  $P'_{n,d}$  is calculated as follow

$$P_{n,d}' = \frac{P_{n,d} + \alpha I_n}{1 + \alpha (N_k - 1)},$$

where  $I_n$  (Neighbors readiness parameter) is a normalized range between [0, 1] which depends on  $q_n$ 

$$I_n = 1 - \frac{q_n}{\sum\limits_{n'=1}^{N_k} q_{n'}}$$

Here,  $q_n$  is the length of the queue of the packages that are sent from node k to its neighbor n and  $\alpha \in [0.2, 0.5]$  is the weight of  $I_n$  in calculation of  $P'_{n,d}$ that has been obtained experimentally.

If one loop is diagnosed, the ant takes out the nodes existing in the loop from stack and selects another node for moving.

3) The forward ants create a stack and place the travel time to each node inside it.

4) When the forward ants reach destination, they give their stack to backward ants and they die.

5) The backward ants take out the data items from the stack and take the route back toward the starting node.

6) The routing table of each visited node is updated [8, 28-32].

7) After getting to the destination, the backward ant

also dies.

## 3. Methodology

In this section, we describe our hybrid method explicitly. In the presented algorithm, like CPANT algorithm [27], we assume that the traffic is classified into 4 categories Shown in **Table 2**.

**Conversational:** such as VOIP or video conference traffic (video conference).

**Streaming:** where a play-out memory (buffer) can mitigate the effects of jitter and where no interaction (action and reaction) takes place; e.g. watching a video stream or listening to a pod-cast.

**Interactive:** (active and reactive), with lower bandwidth requirements such as Web surfing and Web applications.

**Background:** such as Email and large File transfers e.g. FTP (file transfer protocol) or P2P-(Pair to Pair) File sharing.

We introduce these four classes with the notion of color, including Red, Blue, Green and Black.

We use 3 approaches in our method:

1) The ants mark paths throughout the network by depositing pheromones with different colors depending on the suitability of the path for the corresponding traffic classes. For example a path with high bandwidth, low jitter, and low delay is suitable for traffic in the conversational class and would therefore be marked with pheromones of the "Red" color.

2) On the condition that the traffic should be preferably routed with its major color, it can further use paths marked with a color other than its own if no path of appropriate color can be found. For example if all paths are of such high quality that they are all marked with color "Red", then traffic of classes 2, 3, and 4 can also use these "Red-path" (**Figure 1**) [27].

3) Traffic is always sent through the best color path, but when the load of the route increases, its pheromone value is reduced over the time and when it becomes worse than grade two path, traffic automatically move to grade two path.

In the proposed algorithm, we use three types of ants: forward ants, backward ants, and updating ants [24] that play the role of creating cooperation between the neigh-

Table 2. The traffic classes and their differing critical and uncritical requirements.

Class	Name	Critical	Uncritical
1	Conversational	Bandwidth, jitter, delay	
2	Streaming	Bandwidth	Jitter, delay
3	Interactive (world wide web)	Delay	Jitter, (bandwidth)
4	Background (data)	-	Jitter, delay, (bandwidth)

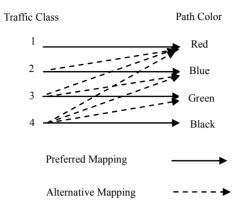


Figure 1. Mapping of different traffic classes and alternative paths for each class.

bors as well as transferring of information. The data structure of this algorithm is similar to AntNet (Routing table  $P_k$  and Table  $M_k$ ).

The stages of hybrid algorithm are as follows:

1) In regular intervals, each node s generates a forward ant and sends it to the destination d. The forward ants mark paths throughout the network by depositing pheromones with different colors depending on the suitability of the path for the corresponding traffic classes.

2) According to AntNet algorithm, the forward ant finds a path to destination *d*, with an approach to prevent forming of loops and arrives at destination. The forward ant maintains in its memory the particulars and the time required to reach the nodes existing on the path.

3) Upon arriving at the destination, the forward ant transfers its memory to the backward ant and dies. The path is also graded and the grade of the route is added to the backward ant.

The color vector for the current link taken. The color values a, b, c, and d are in the interval [0, 1] and are calculated using thresholds which map link attribute values to percentage values of the color. This mapping can be implemented using a simple lookup table in the routing nodes. The ant keeps an ordered list

 $L_{ant} = \{\overline{X}_1, \overline{X}_2, \dots, \overline{X}_n\}$ , of the color vectors  $\overline{X}$  encountered along the way. When the ant reaches to the destination node, the path is graded for all colors by calculating the path grade [27]:

$$\overline{G}_p = \prod_j \overline{X}_j \quad \forall \overline{X}_j \in L_{ant}.$$

4) The backward ant travels back from the same path that taken by the forward ant and updates the table.

5) If the forward ant has taken the distance between node k to destination d in  $\Delta T_{kd}$  and the backward ant in the node k decides that this is a suitable path, *i.e.*  $\Delta T_{kd} < T_{Best_d}$ , the updating ants are generated at the number of neighbors of K and places the ID of node K and  $\Delta T_{kd}$  and  $T_k$  (current time in node k) within their

memory and send them to node k's neighbors  $(n_i)$ 's).

6) When the updating ant enters into the node *i* from the node k, it would calculate the  $\Delta T_{ki}$ ,

$$\Delta T_{ki} = T_i - T_k,$$

where  $T_i$  is the current time in node *i*. If we assume that  $\Delta T_{ik} = \Delta T_{ki}$ , we can estimate the time  $\Delta T_{id}$  required to travel from *i* to *d* through the node *k* by [24]:

$$\Delta T_{id} = \Delta T_{ik} + \Delta T_{kd} \, .$$

Now if  $\Delta T_{id} < T_{Best_d}$ , *i* will change its routing table for the destination *d*. Here, the net time spent, is  $\Delta T_{id}$  and the next node will be *k*. So, in the routing table *i*, for the destination *d*, the possibility of  $P_{kd}$  (possibility of selecting *k* as the next node for traveling to *d*) increases and the possibility of selecting other neighbors of node *i*  $(n \in N, n \neq k)$  decreases [27]:

$$\tau x_{k,d}^{i}\left(t\right) = \tau x_{k,d}^{i}\left(t-1\right) \cdot f_{evap} + g\left(Gx_{p}\right) \cdot f_{evap}$$

Here  $Gx_p$  denotes that element of the path grade vector that matches the color of the pheromone being updated.  $f_{evap}$  denotes the evaporation function and  $g(Gx_p)$ represents the enforcement function and t-1 denoting "the last value".  $\tau x_{k,d}^i$ ,  $Gx_p$  are always used with identical values for x; for example in order to calculate  $\tau A_{k,d}^i$  value of  $GA_p = a^*$  is used.

For all other links that are not placed on the path (meaning an  $i \rightarrow m$  hop, provided m is contrary to k), the value of pheromone will reduce [27]:

$$\tau x_{m,d}^{i}\left(t\right) = \tau x_{m,d}^{i}\left(t-1\right).f_{evap}.$$

Examples of such functions can be in the form of  $f_{evap} = 1 - Gx_p$  and  $g(Gx_p) = Gx_p^k$  that lead to  $0 \le f_{evap} \le 1$  and  $0 \le g(Gx_p) \le 1$  for  $0 \le Gx_p \le 1$ .

#### 4. Experiments and Results

We call our method Modified Colored Pheromones Ant Routing (MCPANT). MCPANT has been simulated by Glomosim 2.03. Considering the randomicity in our simulations, each datum is the mean value of 20 times independent calculations. In this section we want to compare MCPANT with standard AntNet and CPANT Algorithms.

In this simulation, a network topology model with 12 nodes and 17 bidirectional links are used (see **Figure 2**). Every link has three specifications: delay (ms), jitter (ms), bandwidth (mbps). Nodes 0, 1, 2 and 3 are sources of packet traffic generations. The traffic sources are constant bit rate (CBR), each packet length is 512 bytes and the total simulation time is 500 seconds. Each node has an incoming packet buffer with a maximum capacity of 1024. A traffic route should be determined before a traffic flow is going to be sent off at its generating node, and

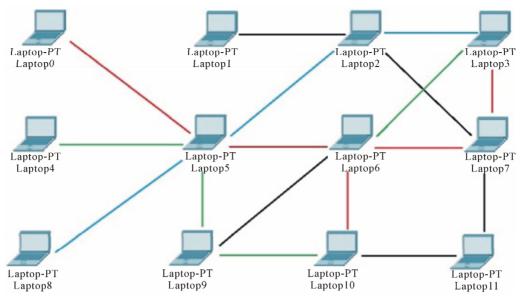


Figure 2. Simulation topology with certain bandwidth, delay and Jitter for each link.

the route will be determined according to the routing tables of nodes. The total number of packets generated from different sources for distinctive destinations are 2000 packets. For more details, see **Table 3**.

## 4.1. Topology

In the simulation process, we have used the topology proposed in [27] and Shown in **Figure 2**.

We have defined 4 traffic classes as hereunder:

1) Conversational traffic (red color) with a bandwidth of 20 Mbps and delay of 1 ms and jitters of 1 ms.

2) Streaming traffic (blue color) with a bandwidth of 20 Mbps and delay of 40 ms and jitters of 8 ms.

3) Interactive (web) traffic (green color) with a bandwidth of 8 Mbps and delay of 1 ms and jitters of 16 ms.

4) Background (data) traffic (black color) with a bandwidth of 4 Mbps and delay of 80 ms and jitters of 32 ms.

In each link, based on the features of the link (bandwidth, delay and jitters in the link), the highest color is chosen from among four traffic classes, while in the standard AntNet, there was only one color of pheromone due to non classification of the traffic.

The delay D, for the path R, has been obtained by

$$D = \sum_{l \in \mathbb{R}} D_l, D_l$$
 is delay on the link *l*.

Also the jitter *J*, for the path *R*, is calculated as follow:

$$J = \sum_{l \in R} J_l, J_l$$
 is jitter on the link *l*.

## 4.2. Metrics of Performance Evaluation

Simulation results of our method are compared with standard AntNet and CPANT based on following criteria.

## 4.2.1. End to End Delay

End to End Delay is the time required for the information packet to travel from the source to final destination.

Table 3. Parameters of the simulation.

Parameters of Simulation				
Simulation-time	500 S			
Seed	3			
Terrain-dimensions	(400, 800)			
Number-of-nodes	12			
Node-placement	FILE			
Mobility	NONE			
Propagation-limit	-111.0			
Propagation-pathloss	Two-Ray			
Noise-figure	10.0			
Temparature	290.0			
Radio-type	Radio-Accnoise			
Radio-frequency	2.4e-9			
Radio-bandwidth	2,000,000			
Radio-rx-type	SNR-Bounded			
Radio-rx-SNR-threshold	10.0			
Radio-tx-power	7.874			
Radio-antenna-gain	0.0			
Radio-rx-sensitivity	-91.0			
Radio-rx-threshold	-81.0			

While the algorithm finds the optimum path, this delay be is kept to minimum. When a change or an error is occurred in the network, the quick convergence of the algorithm through the help of quality of the bandwidth will play a crucial role in mitigating the delay. The delay index will therefore be suitable to compare the speed of convergence of algorithms.

### 4.2.2. Jitter

For some services, the jitters are much more important than the delay itself. In case the algorithm has been delayed for convergence or in case of interference, the jitters will also increase. This index has also been used in comparisons.

#### 4.3. Experiments

Reaching convergence will make the routing tables to be quantified with optimum paths and as a result mitigating the delay and jitters. Hence, if upon the start of each simulation, the data traffic is transmitted throughout the network, we can distinguish through the comparison of the above criteria that which algorithm has reached convergence sooner.

#### The Cumulative Distribution Function (CDF)

Let x be a random variable the Cumulative Distribution Function  $F_x(x)$  is an ascending non-zero and conforming function with a interval range of [0, 1]. This Function shows the possibility that *X* has got smaller quantity than *x*, this means:

$$F_{X}(x) = P(X \le x).$$

In **Figure 3**, the results of delay of standard AntNet algorithm were compared with those of the hybrid algorithm and as you can see the results show that the delay in conversational, streaming and interactive classes is less than in standard AntNet and in the background class, the more time goes ahead, the lower the delay will be than standard AntNet.

**Figure 4** shows a comparison of the results of delay in CPANT algorithm and the hybrid algorithm, and due to the existence of updating ants in the intermediate nodes, the speed of convergence has increased and the results of delay have improved in comparison to CPANT algorithm. The overall delay average of the network is decreased 5.23% in comparison to CPANT algorithm because of existence of updating ants.

In **Figure 5**, a comparison has been made between the results of the jitter in standard AntNet algorithm and hybrid algorithm; the results obtained from hybrid algorithm are better than the standard AntNet algorithm and in comparison of the traffic classes, it is also observed that taking the features of the classes into consideration, Class conversational has got the lowest changes and Classes streaming, web and data stand in the next ranks. In **Figure 6**, considering the cumulative distribution

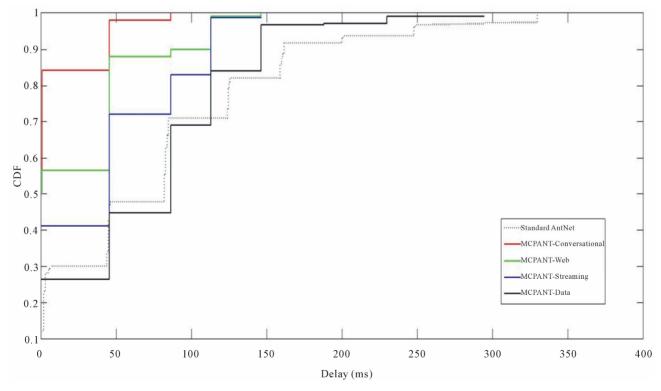


Figure 3. Comparison of the delay in standard AntNet algorithm and hybrid algorithm (MCPANT).

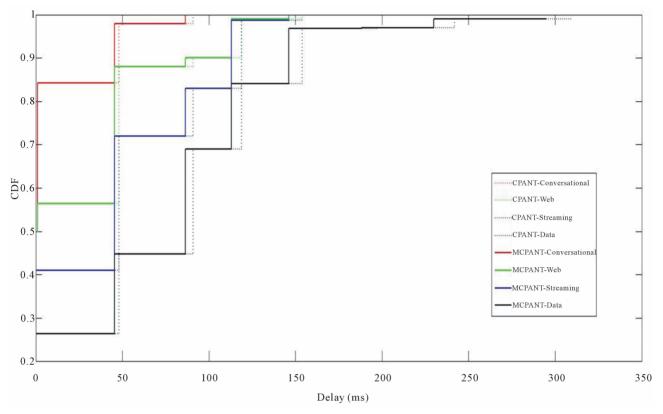


Figure 4. Comparison of the delay in CPANT algorithm and hybrid algorithm (MCPANT).

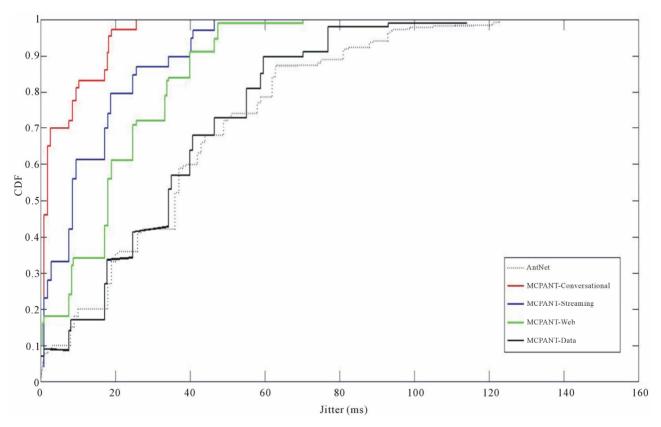


Figure 5. Comparison of the jitter in standard AntNet algorithm and hybrid algorithm (MCPANT).

14

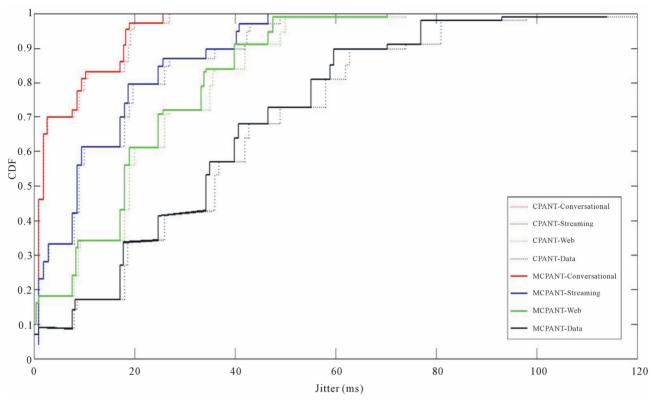


Figure 6. Comparison of the jitter in CPANT Algorithm and hybrid Algorithm (MCPANT).

Routing Algorithm Standard Criteria	Standard AntNet	Modified AntNet (Helping Ant)	CPANT	MCPANT
AVG. End-to-End Delay (ms)	14.28	13.56	1.43	1.36
			2.01	1.92
			2.32	2.1
			3.81	3.6
AVG. Throughput (Packet/Sec)	215	236	262	297
Packet Drop Ratio (%)	6	5.54	5.6	5.13
Overhead (%)	6	8	6	8

Table 4. Experimental results obtained from 20 simulations of the four competing algorithms.

function (CDF), the results of jitter of CPANT algorithm and the hybrid algorithm have been compared and taking the jitter in alternative classes into consideration and although there are updating ants in the intermediate nodes, the results of the jitter 5/8% have improved compared to CPANT algorithm. As an example, jitter of blue class is decreased from 19.7 ms in CPANT algorithm to 18.715 ms in the hybrid algorithm.

The experimental results for twenty different simulations of these four algorithms are summarized in **Table 4**.

### 5. Conclusion and Future Research

In this paper, a hybrid method is presented that uses helping ants for updating the routing tables for speeding up the convergence. In the proposed method, for categorizing the needs traffic classification by the help of colored pheromones.

The simulation of different status of the network and its comparison with the standard AntNet algorithm and CPANT algorithm reveals that the use of hybrid method has both made the convergence speed faster and mitigated the delay and jitter. This is while even with reducing the generation of agents, the hybrid algorithm with a less overload has got a better performance than the compared algorithms. In future we will design and simulate the existing algorithms in a fuzzy state in order to increase the flexibility of the algorithm.

### REFERENCES

- I. F. Akyildiz and I. H. Kasimoglu, "Wireless Sensor and Actor Networks: Research Challenges," *Ad Hoc Networks*, Vol. 2, No. 4, 2004, pp. 351-367. doi:10.1016/j.adhoc.2004.04.003
- [2] I. F. Akyildiz, W. Su, Y. Sankarasubramaniam and E. Cayirci, "Wireless Sensor Networks: A Survey," *Computer Networks*, Vol. 38, No. 4, 2002, pp. 393-422. doi:10.1016/S1389-1286(01)00302-4
- [3] I. F. Akyildiz, X. Wang and W. Wang, "Wireless Mesh Networks: A Survey," Broadband and Wireless Networking (BWN) Lab, School of Electrical and Computer Engineering, Georgia Institute of Technology, Atlanta, 2005, pp. 445-487.
- [4] S. M. Alamouti, "A Simple Transmit Diversity Technique for Wireless Communications," *IEEE Journal on Selected Areas in Communications*, Vol. 16, No. 8, 1998, pp. 1451-1458.
- [5] M. Alicherry, R. Bhatia and L. Li, "Joint Channel Assignment and Routing for Throughput Optimization in Multi-Radio Wireless Mesh Networks," *Proceedings of* the 11th Annual International Conference on Mobile Computing and Networking, ACM, New York, 28 August-2 September 2005, pp. 58-72.
- [6] R. Draves, J. Padhye and B. Zill, "Routing in Multi-Radio, Multi-Hop Wireless Mesh Networks," ACM Annual International Conference on Mobile Computing and Networking, 26 September-1 October 2004, pp. 114-128.
- [7] R. Bhatia and M. Kodialam, "On Power Efficient Communication over Multi-Hop Wireless Networks: Joint Routing, Scheduling, and Power Control," *IEEE Annual Conference on Computer Communications*, Hong Kong, 7-11 March 2004, pp. 1457-1466.
- [8] C. Blum, "Ant Colony Optimization: Introduction and Recent Trends," Universitat Politècnica de Catalunya, Barcelona, 2005, pp. 353-373.
- [9] G. Held, "Wireless Mesh Networks," Taylor & Francis Ltd., London, 2008.
- [10] O. Hussein, T. Saadawi and M. Lee, "Probability Routing Algorithm for Mobile Ad Hoc Networks," *Journalon Selected Areas in Communications*, Vol. 23, No. 12, 2005, pp. 2248-2259. doi:10.1109/JSAC.2005.857205
- [11] T. H. Labella and F. Dressler, "A Bio-Inspired Architecture for Division of Labour in SANETs," *Proceedings of* 1st IEEE/ACM International Conference on Bio-Inspired Models of Network, Information and Computing Systems, Cavalese, 2006.
- [12] B. Ko, V. Misra, J. Padhye and D. Rubenstein, "Distributed Channel Assignment in Multi-Radio 802.11 Mesh Networks," WCNC 2007: Proceedings of the IEEE Wire-

- [13] M. Kodialam and T. Nandagopal, "Characterizing the Capacity Region in Multi-Radio Multi-Channel Wireless Mesh Networks," *Proceedings of the 11th Annual International Conference on Mobile Computing and Networking*, ACM, New York, 11-15 March 2005, pp. 73-87.
- [14] A. M. Rad and V. Wong, "Joint Optimal Channel Assignment and Congestion Control for Multi-Channel Wireless Mesh Networks," *Proceedings of the IEEE International Conference on Communications*, Vol. 5, 2006, pp. 1984-1989.
- [15] K. Ramachandran, E. Belding, K. Almeroth and M. Buddhikot, "Interference-Aware Channel Assignment in Multi-Radio Wireless Mesh Networks," *Proceedings of the 25th IEEE International Conference on Computer Communications*, 23-29 April 2006, pp. 1-12. doi:10.1109/INFOCOM.2006.177
- [16] A. Raniwala, K. Gopalan and T. Chiueh, "Centralized Channel Assignment and Routing Algorithms for Multi-Channel Wireless Mesh Networks," *SIGMOBILE Mobile Computing and Communications Review*, Vol. 8, No. 2, 2004, pp. 50-65. <u>doi:10.1145/997122.997130</u>
- [17] J. Tang, G. Xue and W. Zhang, "Interference-Aware Topology Control and Qos Routing in Multi-Channel Wireless Mesh Networks," *Proceedings of the 6th ACM International Symposium on Mobile adhoc Networking and Computing*, ACM, New York, 25-28 May 2005, pp. 68-77.
- [18] Y. Zhang, J. Zheng and H. Hu, "Security in Wireless Mesh Networks," Wireless Networks and Mobile Communications, Taylor & Francis, London, 2008, p. 552.
- [19] H. Wu, F. Yang, K. Tan, J. Chen, Q. Zhang and Z. Zhang, "Distributed Channel Assignment and Routing in Multiradio Multichannel Multihop Wireless Networks," *IEEE Journal on Selected Areas in Communications*, Vol. 24, No. 11, 2006, pp. 1972-1983. doi:10.1109/JSAC.2006.881638
- [20] R. Schoonderwoerd, O. Holland and J. Bruten, "Ant-Like Agent for Load Balancing in Telecommunications Network," *Proceedings of the 1st International Conference* on Autonomous Agents, 5-8 February 1997, pp. 209-216. doi:10.1145/267658.267718
- [21] G. D. Caro and M. Dorigo, "AntNet: Distributed Stigmergetic Control for Communication Networks," *Journal* of Artificial Intelligence Research, Vol. 9, No. 1, 1998, pp. 317-365.
- [22] G. D. Caro and M. Dorigo, "Mobile Agent for Adaptive Routing," *Proceedings of 31st Hawaii International Conference on System Science*, Kona, 5-8 January 1998.
- [23] D. Câmara and A. A. F. Loureiro, "A Novel Routing Algorithm for Hoc Networks," *Baltzer Journal of Telecommunications Systems*, Vol. 18, No. 1-3, 2001, pp. 85-100.
- [24] A. Soltani, M. R. Akbarzadeh and M. Naghibzadeh, "Helping Ants for Adaptive Network Routing," *Journal* of the Franklin Institute, Vol. 343, No. 4-5, 2006, pp.

389-403.

- [25] C. Bertelle, A. Dutot, F. Guinand and D. Olivier, "Distribution of Agent based Simulation with Colored Ant Algorithm," *Proceedings of 14th European Simulation Symposium*, Dresden, 2002.
- [26] K. M. Sim and W. H. Sun, "Ant Colony Optimization for Routing and Load-Balancing: Survey and New Directions," *IEEE Transactions on Systems, Man, and Cybernetics, Part A: Systems and Humans*, Vol. 33, No. 5, 2003, pp. 325-336.
- [27] M. Umlauft and W. Elmenreich, "QoS-Aware Ant Routing with Colored Pheromones in Wireless Mesh Networks," 4th International IEEE Conference "Intelligent Systems", Vol. 3, 2008, pp. 1060-1064.
- [28] M. Dorigo and C. Blum, "Ant Colony Optimization Theory: A Survey," Université Libre de Bruxelles, Brussels, Belgium, Universitat Politècnica de Catalunya, Barcelona,

2005, pp. 243-278.

- [29] M. Dorigo, M. Birattari and T. Stutzle, "Ant Colony Optimization," *IEEE Computational Intelligence Magazine*, Vol. 1, No. 4, 2006, pp. 28-39.
- [30] M. Gunes, U. Sorges and I. Bouazizi, "ARA—The Ant-Colony Based Routing Algorithm for MANETs," *Parallel Processing Workshops*, Vancouver, 20-23 August 2002, pp. 79-85.
- [31] D. Pan and B. Xing, "An Ant Routing Algorithm for Wireless Mesh Network," *Proceedings of the 7th World Congress on Intelligent Control and Automation*, Chongqing, 25-27 June 2008, pp. 4595-4599.
- [32] D. Pan, Y. Xue and L. Zhan, "A Multicast Wireless Mesh Network Routing Algorithm With Ant Colony Optimization," *Proceedings of the 2008 International Conference* on Wavelet Analysis and Pattern Recognition, Hong Kong, 30-31 August 2008, pp. 744-748.