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Smart Data Packet Ad Hoc Routing Protocol

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Abstract—This paper introduces a smart data packet routing protocol (SMART) based on swarm technology for mobile ad hoc networks. The main challenge facing a routing protocol is to cope with the dynamic environment of mobile ad hoc networks. The problem of finding best route between communication end points in such networks is an NP problem. Swarm algorithm is one of the methods used solve such a problem. However, copping with the dynamic environment will demand the use of a lot of training iterations. We present a new infrastructure where data packets are smart enough to guide themselves through best available route in the network. This approach uses distributed swarm learning approach which will minimize convergence time by using smart data packets. This will decrease the number of control packets in the network as well as it provides continues learning which in turn provides better reaction to changes in the network environment. The learning information is distributed throughout the nodes of the network. This information can be used and updated by successive packets in order to maintain and find better routes. This protocol is a hybrid Ant Colony Optimization (ACO) and River formation dynamics (RFD) swarm algorithms protocol. ACO is used to set up multi-path routes to destination at the initialization, while RFD mainly used as a base algorithm for the routing protocol. RFD offers many advantages toward implementing this approach. The main two reasons of using RFD are the small amount of information that required to be added to the packets (12 bytes in our approach) and the main idea of the RFD algorithm which is based on one kind of agent called drop that moves from source to

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destination only. This will eliminate the need of feedback packets to update the network and offers a suitable solution to change data packet into smart packets. Simulation results shows improvement in the throughput and reduction in end to end delay and jitter compared to AODV and AntHocNet protocols.

Index Terms— Ad hoc networks, RFD optimization, Routing protocols.

1. INTRODUCTION

The absence of infrastructure in Mobile ad hoc network (MANET) and the dynamic and continuously changing network topology compose real challenges to routing algorithms [1]. Routing in such a network is considered as an optimization process of locating the optimal paths between sources and destinations. A number of algorithms have been proposed to address routing in ad hoc networks [2-6].

Routing protocols for mobile ad hoc can be generally classified into three types. Firstly, proactive routing protocols, which also can be called as table-driven routing protocols, such as Optimized Link State Routing protocol (OLSR) [7], and Destination-Sequenced Distance Vector (DSDV) routing protocol [8]. Secondly, reactive routing protocols, which also can be called as on-demand routing protocols, such as Ad hoc On Demand Distance Vector (AODV) routing protocol [9], and Dynamic Source Routing (DSR) protocol [10]. Finally, hybrid routing protocols such as Zone Routing Protocol (ZRP) [11], and Hybrid On-demand Distance Vector Multi-path (HODVM) routing protocol [12].

Swarm intelligence (SI) has widely been used to address the problem of finding optimal routes [5]. Cognitive Packet Network (CPN) is another approach to solve routing problem which is based on on-line sensing and monitoring of network Quality of Service (QoS).[13]

Designing a protocol with two or more QoS constraints is known to be a NP-complete problem [14]. As problem size increases, the computational complexity of a problem increases, and more time and computation power are required to solve such a problem. Accordingly, finding shortest path as well as minimizing delay and avoid congested nodes is an NP-complete problem. Swarm intelligence has been adapted to solve routing problems due to their efficiency and distributed approach. However, to solve complex problem using swarm algorithms, the number of iterations required will be proportional to problem complexity. As the network become more dynamic, the topology will change faster, and more swarm agents are needed to cope and quickly adapt with this changes in the network. In highly dynamic environment, more agents are required to detect link qualities in order to find best route. Increasing the rate of agent, in order to quickly adapt to the changes, will consume more resources. Moreover, each agent will require a feedback agent to adapt learned parameters which will increase resource consumption.

This paper will propose a smart data packet protocol for mobile ad hoc network using RFD and ACO algorithms. While the main idea is based on the RFD algorithm, ACO algorithm is used to address the bottleneck of slow convergence rate of the RFD algorithm in the starting of learning procedure [15]. The using of ACO algorithm will build a multipath route from source to destination at the beginning of route setup. We introduce the use of data packet in the learning process of swarm algorithm. Rather than increasing the number of agents, we use data packets which act like drop agents. We call these packets "smart data packets" because they behave in smart way by avoiding congested nodes and they incorporate in the learning process. Moreover, we call them smart to distinguish them from ordinary packets (packets that are not acting like drops). As these data packet moves in the network, they adapt altitude tables, and and the network learn about its environment. Using smart data packet will accelerate the sensing and reaction toward network parameters change.

ACO has been widely used for solving routing problem. In general, in ant based routing protocol, a node generate forward ant packets to find the destination. Ant packets move around the network in a random

walk to find the destination. This movement is based on some stochastic probability function. When the destination is found, a backward ant is send from the destination toward the source. As the backward ants move back to the source, they update the pheromone intensity on the links between the nodes. The path with higher pheromone intensity will attract more ants and data packets. After a period of time, the optimal route, depending on the optimization parameters, will be become attractive and will be chosen by the data packets.

River formation dynamics is a subset of swarm intelligence. It reflects how raindrops on highlands join together to form rivers [16]. These rivers tend to take shortest path to the sea. Implementing the RFD algorithm in ad hoc routing protocols provides many advantages. First of all, as there is no backward agent in the RFD algorithm, it will decrease the total number of control packets in the network. Another advantage is the simplicity of the algorithm; especially it relates altitudes to nodes rather than links. As generally the number of nodes is usually less than the number of links in a network. This minimizes the resource usage. More advantages comes from the fact that the implemented RFD based protocols are using promiscuous communication mode, so the learning process is not local but all neighbor will be affected by the learning process. As drops are moving in the network, they update the altitude of the nodes. The drop carries recent node altitude which will be detected by neighboring nodes. All neighbor nodes will update the corresponding node altitude in their tables. In other words, one change in a node's altitude which announced by one drop packet is corresponding to changes of all links between that node and all its neighbors. A farther advantage is since RFD uses just drops, which element the need for backward agents and RFD drops adapts network parameters while they are moving from source(s) to destination(s), this will offer the opportunity to use data packets and make them act like drops. This allows data packets to guide themselves and contribute in the learning process. Other protocols usually require backward agents to adapt routing parameters. With other protocol, assigning a backward agent for each data packet will exhaust the network. Finally, the amount of control information appended to data packets is small, which make it easy to integrate the information into the data packets.

This paper is organized as follows. In the next section, related works are given briefly. Section three will explain the main idea of the RFD algorithm. Section four will explain the proposed protocol. The results and implementation are given in section five. Finally, a conclusion is given in section six.

2. RELATED WORKS

Routing algorithm is an optimization process that tries to maximize network performance while minimizing costs. In [9] AODV introduced to solve routing problem. AODV is one of the most popular classical routing protocols for mobile ad hoc networks. Whenever a node needs to send data to a destination, and it does not have the valid route to destination, it broadcast a Route Request (RREQ) message to find the destination. Upon receiving RREQ, Route Replay (RREP) message is send back to the source. AODV in its original form uses hello message to periodically update its neighbor nodes availability. Link breakage could be detected if unsuccessful packet transmission occurs or missing hello message. In case of link failure the node send back a Route Error (RERR) to the source in order to search for new route. DSR [10] is another type of on demand ad hoc routing protocol which uses dynamic source routing. Unlike hop to hop routing, source routing protocol adds the complete route path to the packet which gives the source node complete control on how the packet moves in the network. Ad hoc On demand Multipath Distance Vector (AOMDV) [17] is a multipath routing protocol which can deal better with mobility than AODV protocol. AOMDV is an enhancement to AODV where multiple paths are created at route discovery process. These paths are disjoints and loop free. AOMDV offers better throughput than AODV regarding and less control packet overhead.

OLSR [7] is a table driven routing protocol. Each node in OLSR uses hello messages to discover its two hop neighbors using the information carried by hello messages. Each node can select its multipoint relay MPR nodes. MPR nodes are a set of nodes that allow a node to communicate with all its two hop nodes through these MPR nodes. Each node maintains a table for its neighbors (neighbor table) as well as another table that contains addresses of it neighbor nodes that have selected it as a MPR (MPR selector table).

Topology control (TC) message are broadcasted periodically and used to build routing tables. DSDV [8] is another table driven routing protocol where every node keeps a set of distances to every destination throughout its neighbors. The distance metric can be the number of hops or the end-to-end delay. In order to keep the distance vector tables up to date, each node periodically broadcasts an update of its shortest routes to its neighbors.

In [18] the authors analyze the performance of different types of routing protocols used in Wireless Networked Robotics (WNR). Different scenarios have been proposed to identify the features that affect the performance of traditional ad hoc routing protocols. The study shows both node capacity and traffic have the major impact on the performance of routing protocols in WNR. Finally, the study shows that in average the AODV performance could be considered better than OLSR, DSR, and DSDV.

In [12] HODVM routing protocol for Spatial Wireless Ad Hoc (SWAH) networks is proposed. Spatial Wireless ad hoc network consists of both mobile and static nodes. Two different protocols have been adapted to work with backboned network and non-backboned network. Static routing is used for static nodes, while AODV routing protocol is used for dynamic nodes. Moreover a node behavior distinguishing algorithm is used to select multiple routes.

Adaptive Neuro-Fuzzy Inference System (ANFIS) has been used by [19] to select a single destination (server) from a group members belonging to anycast group in mobile ad hoc networks. The protocol uses three kinds of agents in order to cope with the QoS required. These agents are: static anycast manager agent, static optimization agent, and mobile anycast route creation agent. Moreover the protocol tries to select stable routes.

Swarm intelligence is well known optimization algorithm which inspired from the social behavior of insects and other animals and used to solve optimization problem. Ant Colony Optimization is one of swarm technology algorithm that has been used to solving routing problems [20]. Ant Based Control protocol (ABC) [21] is considered the first SI routing protocol for telecommunication networks. ABC protocol addresses load balance problem using ant agents. The routing protocol is proposed to work on

circuit switched networks. The ants move in one direction from sources to other nodes. As ant moves they deposit pheromones which will eventually guide data packets. ANTNET [22] is a routing protocol for packet switch networks. ANTNET uses forward and backward ants based on ACO. ANTNET uses distance vector for data routing, while it uses source routing for control packets. Thus it will introduce high overhead especially in large networks. Ant-AODV [23] is a hybrid routing protocol that combines the idea of AODV and ant-based algorithm. Ant-AODV reduces the end to end delay but it increases the overhead of route discovery and maintenance. Ant Colony Based Routing protocol (ARA) [24] is using distance vector routing and supports multipath. It is a reactive routing protocol. The protocol tries to limit the overhead caused by ants but it losses the proactive feature of ants algorithms. AntHocNet [25] is a multipath hybrid routing protocol that uses source routing principles combined with ACO. Ants in this protocol compare both the travel time and hop count with previous visited ants and only broadcast if it is better. Certain concentrate have been added to decide if ants are going to be broadcasted or unicasted in the network. AntHocNet protocol features automatic load balancing as backward ants take into account the delay in each hops. Hello messages have been used to discover neighbor nodes and defuse pheromones. Link failure is treated using local repair technique. In case of route failure, the node even forward the packet to next best available neighbor node or it will try to locally repair the route by broadcasting route repair ant if no more links is available in its route table. On both cases the node informs its neighbor about link failure. This local repair technique usually will not lead to optimal solution rather than it only finds another path to the destination.

In [26], Hybrid Routing Algorithm based on ant colony and ZHLS routing protocol for MANET (HRAZHLS) is proposed. HRAZHLS is based on the Zone based Hierarchical Link State (ZHLS) protocol. The protocol uses reactive routing outside the zones and used to update its interzone routing table (InterRT). On the other hand, it uses proactive routing inside the zones to update its intrazone routing table (IntraRT). Different types of ant are used in the protocol like internal forward ant, external forward ant,

backward ant, notification ant and error ant. Internal forward ants are used to build the proactive tables, while external forward ants are used to build the reactive routing tables.

BeeAdHoc [27] mimics the beehive in nature. It is relatively a simple algorithm that makes use of a reactive strategy for agent launching and of source-routing to forward packets. There are two types of bee agents in the network, short and long distance bee agents. These agents are responsible of exploring the network and evaluate the quality of their paths in order to update node routing table.

Cognitive Packet Network (CPN) introduced the use of intelligent packets where the capabilities for routing and control have been moved towards the packets themselves. In CPN, Random Neural Networks (RNNs) has been used in order to make routing decision. CPN contains three types of packets: Smart Packets (SPs), Dumb Packets (DPs) and Acknowledgement (ACKs). Packets contain extra fields for Cognitive Map (CM) and Executable Code. CPN has been used as routing protocol for ad hoc network in [28, 29].

In [30] the authors present a multiple path approach for CPN in order to perform load balance among all network nodes. The algorithm carried out in two steps. The first step collect the information about available multipath in the network then a Hopfield neural network algorithm is used to refine and balance the distribution of packet around the network.

CPN suffers from high overhead as the amount of control information added to the packet is high. CPN uses Random Neural Networks (RNN) which adds more computation and resource usage to the network, as well as vast amount of information (neural networks weights) should be carried by the packets. CPN protocol infrastructure is completely different from the well-known layered approach of routing infrastructure. These and other factors led to less interest in CPN.

RFD is a swarm algorithm and has been applied in many combinatorial optimization problems such as the asymmetric traveling salesman problem [16, 31], Optimal Quality-Investment Tree problem [32], minimum spanning Tree Problem [33], and others [34-36]. The RFD algorithm mainly uses one kind of

agents which is called drops. Drops moves only from sources to destinations. The RFD algorithm is a competitor to ACO algorithm and has shown to perform better than ACO in many applications [32, 35].

Our proposed protocol is based on the RFD algorithm. It shares some common feature with the above presented protocols. It is a hybrid routing protocol. It uses reactive route setup whenever a route is not available. At the same time it uses hello messages to defuse topology information. Like AntHocNet and other swarm protocols, SMART protocol uses agents to search for best path. However, SMART protocol uses data packet to act like drop agents. Unlike other swarm protocols, there is no need for backward agents. Like CPN, routing decisions are influenced by the information carried by data packets. SMART protocol distributes the learned information around the network instead of carrying it within the packets. Only part of the information (altitudes) which reflect the change of network state will be carried by the packets. Moreover SMART protocol uses distributed learning which is more efficient than local learning algorithm proposed by CPN.

3. RIVER FORMATION DYNAMICS ALGORITHM

The process of the RFD algorithm starts with initializing the nodes altitude to predetermined positive value, which reflect flat surface at the beginning. The only exception is the goal point or destination which will have an altitude value of zero. Destination is considered as sea where the drops should end at. Drops are generated at the source (sources). At the beginning as all the nodes have same altitude value, the drops will spread around the flat environment. When some drops find the destination, they fall into it. As they fall, the nodes altitudes will be eroded. This erosion will create a down slop and throughout many training cycles the slop will be propagated backward to the source. Drops move according to the following random probability selection.[16, 33]

$$P_{k}(t, j) = \begin{cases} \frac{decreasingGradient(t, j)}{\sum_{l \in V_{k}(t)} decreasingGradient(t, l)}} & \text{if } j \in V_{k}(t) \\ 0 & \text{if } j \notin V_{k}(t) \end{cases}$$
(1)

where $\mathbb{P}_k(i,j)$ is the probability of drop *k* at node *i* to select node *j*. V_k is a set of neighbors nodes that can be visited by the drop from node *k*. *decreasingGradient*(*i*,*j*) represents the negative gradient between nodes *i* and *j*, which is defined as follows:

$$decreasingGradient(l, f) = \frac{altitude(Q-altitude(f))}{distance(l, f)}$$
(2)

where altitude(x) is the altitude of the node x and distance(i,j) is the length of the edge connecting node *i* and node *j*. At the beginning of the algorithm, all nodes have the same altitude, and the sum of the *decreasing Gradient* is also zero. RFD algorithm suggests giving a special treatment to flat gradients, where the probability that a drop moves through an edge with zero gradient is set to some (non-null) value. This enables drops to spread around a flat environment, which is mandatory, in particular, at the beginning of the algorithm.

When a drop moves off a node to lower altitude, the node that the drop moved from will be eroded. The amount of erosion is proportional to the difference between the altitudes of two nodes.

$$erosion(1) = \propto (altitude(1) - altitude(1))$$
(3)

Another process which follows the erosion is sediments deposit. There are two type of sediment depositing, first is the regular periodic addition of sediment to all nodes at constant rate and the second one is by drops as they move through the network. The amount of sediments that a drop carries throughout its path from source to that node is accumulative. The amount of sediments that will be deposited at each node is proportional to the amount that the drop carries.

sedimention $\equiv \beta * carried_sediment$

(4)

Finally, the path from source to destination is analyzed and the stop condition is checked. If the quality is not good the procedure of drop sending will be repeated until satisfaction quality is reached or when a maximum number of iteration is reached.

The RFD algorithm shares some aspect with ACO algorithms; the main difference is in the RFD altitude values are assigned to the nodes themselves while in ACO the pheromone values are assigned to the links between the nodes. The RFD algorithm could be considered as the gradient version of ACO [15, 35]. However, the RFD algorithm has many advantages over ACO algorithm. The RFD algorithm converges to better solutions when compared to ACO algorithm [32, 35]. Unlike ACO algorithm, in the RFD algorithm there is no possibility for local cycles. ACO algorithm suffers from local cycles. In order to prevent local cycles in ACO, extra techniques are used like buffering visited nodes which has been used in most source routing ACO based protocols. Source routing protocols suffer from both bandwidth and hop-count drawback.

The RFD algorithm is more adaptable to changes. Moreover, the RFD algorithm has two different sedimentation processes. The first is periodical, which is similar to the process of forgetting in ACO. The second one, which gives advantage to RFD, is the way that RFD tends to cumulate sediment in local valleys especially if a dead end is reached. In which case the drop stops and drops all of its sediment in that node causing its altitude to increase which makes it undesirable for other drops to follow. This way of acting is considered as punishment for bad routes and the probability of other drops to select this insufficient route will decreases. In spite of all the advantages of the RFD algorithm, it has a main drawback. The RFD algorithm suffers from slow convergence rate at the beginning of training cycles [15, 33, 34].

To show the slow convergence of the RFD algorithm, a network consisting of a grid of 5 by 10 nodes is used. Drops are sent from node S (node number 1) to node D as shown in Figure 1. First of all, in order to test the number of drops needed to create a path, we forced the drops to move along the shortest path in order to show the delay in path creation. A path is created whenever there is a decreasing slop from the source to destination. Figure 2 shows the number of drops required to create a path. It shows that at least 8 drops are needed to create a path. This is equal to number of nodes between the source and the destination. When designing a routing algorithm using the RFD algorithm, special care should be giving to the starting up mechanism. That is why we have introduced the use of hybrid and RFD and ant protocol in the beginning of the learning process to overcome this problem.

Figure 3.a shows the effect of variable \propto on the nodes altitude curvature for the shortest path between the source and the destination. The value of β is set to 0.5. It is clear that low value will result in slower learning at nodes close to the source. Figure 3.b shows the effect of β on the curve created by the RFD algorithm for the same network and same number of drops and we set \propto to 0.5. We can see that the curve become concave for low values and as the value of β increases it converted to convex curve.

A surface that is almost flat near the source increases the probability of drops to move in direction opposite to the direction of the destination. Moreover, when the surface near the destination becomes flatter, it gives the opportunity to many nodes to deliver the drop to the destination.

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Figure 1. Network topology



Figure 2. Effect of number of drops on nodes altitude



4. SMART DATA PACKET AD HOC ROUTING PROTOCOL

The proposed protocol is a hybrid routing protocol. Hello message has been used to discover neighbor nodes as well as to defuse information around the network. At the same time, when a node requires a connection to any other node, it starts a reactive route procedure through broadcasting ant-drop messages around the network. Smart data protocol is implemented in network layer and uses promiscuous

communication mode to monitor neighbors' packets in order to update its routing tables. Each node will contain an altitude table where the altitudes of other nodes are stored. Another table is used to store the node itself altitudes towards other nodes. Based on the RFD algorithm the selection of next node, regarding a specific final destination, is proportional to altitude differences between the node and its neighbors. The lower the altitude of the next node, the higher the probability to be selected. Drop packets are routed stochastically over different paths using altitude tables. Smart data packets are routed using restricted stochastic algorithm where they can guide themselves and contribute in the learning process. Non smart data packets are routed using minimum altitude as minimum represents the best discovered path. Drops are always sent from source node to destination throw-out the whale data session in order to maintain and monitor the route as well as to search for better route.

Nodes can communicate and share information directly with neighbors and indirectly with other nodes that are out of its transmission range using the diffusion method which is implemented by hello messages.

The route table is represented as altitude table as shown in table 1, where N_i represents neighbor node i. T_i represents time delay of node i. and ALT_{ij} is the altitude to destination j through node i (node i altitude to destination j)

	Table 1: Altitude table								
			Destinations						
0	Neighbors	Time delay	D ₁	D ₂	D ₃		\mathbf{D}_k		
	N_1	T_1	Alt ₁₁	Alt ₁₂	Alt ₁₃		Alt _{1k}		
	N_2	T_2	Alt ₂₁	Alt ₂₂	Alt ₂₃		Alt_{2k}		
	N ₃	T ₃	Alt ₃₁	Alt ₃₂	Alt ₃₃		Alt _{3k}		
	N_j	T_j	Alt _{j1}	Alt _{j2}	Alt _{j3}		Alt _{jk}		

Table 2: My_	_altitudes	table
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destination	My altitude
D_1	Alt ₁
D_2	Alt ₂
D_{k-1}	Alt _{k-1}
D_k	Alt _k

Another important table is my_altitudes table shown in table 2. This table contains the node itself altitudes to other destinations.

4.1 Route setup

Route discovery starts by checking altitude table for the destination. If the source node has no information about the route to the destination, it starts broadcasting ant-drop messages. When an intermediate node receives the ant-drop, it will either unicast or broadcast it. If the intermediate node does not has a route to the destination it will broadcast it, otherwise it will unicast it. In unicast mode, instead of using links' pheromone intensities to route packets, nodes' altitudes are used to select next node. The probability selecting of next node is based on equation 1. The decreasing gradient is calculated according to the following

$$decreasingGradient_{d}(f,k) = \frac{aititude_{d}(f) - aititude_{d}(k)}{T(k)}$$
(5)

where T(k) is the average time that node k needs to send a packet (well be explained in the next subsection), *altitude_d(j)* is altitude of the node *j* toward destination *d*.

Due to broadcasting, the number of ant-drops will increased and will build a multipath from the source to the destination. Ant-drop packet contains a field called travel time. When a node selects the next node it

adds the corresponding next node expected time delay from it altitude table to this field. This accumulated value will reflect the expected time to travel from the source node to that node. If a node receives another copy of the same ant-drop packet, having the same sequence number, it will forward it only if the number of hops or the expected travel time is less than the previous forwarded ant-drop packet. This selective approach will reduce the overall number of ant-drops in the network which decrease the overhead by removing unpromising ant-drop packets from the network. Another condition on forwarding a duplicate copy of the same ant-drop messages is when the first hop that taken by ant-drop is different from the previous ones. This will allow us to build sufficiently multiple disjoint paths which will add more protection against link failure [25, 37].

When an ant-drop reaches the destination, it will be converted to a backward ant-drop. This backward ant-drop moves backward from destination to source depending on the recorded route in the ant-drop received. This backward ant-drop message will update the altitudes table (my_altitudes) which reflects nodes altitude to that destination. Intermediate node will calculate updating factor as following

$$dt_{f} = \frac{1}{\frac{T_{raute}^{f} + t_{hap} * h_{f}}{2}}$$
(6)

where T_{reuse}^{j} is the total accumulated route time by backward ant-drop while traveling back from destination to node *j*. t_{hop} is a parameter used to represent time required to send a packet in unloaded condition which is set to 3 ms [25]. h_j is number of hops from destination to node *j*.

The proposed protocol is mainly based on the RFD algorithm, so the idea of ant colony optimization is adapted to modify altitudes (nodes altitudes) rather than links weights (pheromones intensity). The backward ant-drop will adopt altitude as following

$$altitude_d(f) = v * altitude_d(f) - (1 - v) * dt_f \quad ve[0, 1]$$
(7)

where $altitude_d(j)$ is altitude of the node j (receiver node) toward destination node d. v is the adaptation factor. Choosing small values for v will lead to rapid change of altitudes and forget its learning information while larger values lead to more smother changes. Value of 0.9 is chosen.

Each node saves a copy of its altitude before changing and updating by equation 7. Each time a copy of backward ant-drop passes through a node, equation 7 is computed for the original saved value. Only the best value is conducted.

When the source node receives backward ant-drop packet, it starts sending data packets.

4.2 Smart data packets and drops routing

In order to find better paths and maintain the route, drop packets are sent throughout data session. Data packets are being used to act like drops and called smart packets. As long as data packet travels from source to destination it is better to use these data packets to reinforce the learning process. This makes data packets detect congested nodes and update altitudes. Therefor successive data packet will select different route as the altitudes change. Smart data packets act like drops; they search and react to network condition as well as they contribute in the learning process of the network. The learning information is stored in the altitude tables throughout the network. Although drops and smart packets act the same way, smart data packet are routed in in restricted way. This minimizes the latency as well as decreases the probability of data packet being sent far away from the destination. At the same time drops will continue discovering other parts of the network.

The rate of drops is set to one drop per 0.5 second. The drops are propagated according to the gradient probability function in equations 5 and 1. Using time delay parameter will reduce the probability of selecting congested nodes. The more congested the node, the more it is not preferred to be selected. The higher value T(k) is representing higher distance according to the RFD algorithm and the node is not preferable to be next forwarding node.

Each drop packet contains a field that represents the recent altitude of sender node. Drops change node altitudes by the process of eroding and adding sediments. The node new altitude to the final destination of drop is attached to the drop packet. All neighbor nodes will update their tables according to the new altitude carried by the drop packet. This is acts like a distributed learning procedure where one change affects a group of nodes. Each node also monitors its sent packets. Unlike ant based algorithms where an ant updates

specific link pheromone intensity that is moving along, one drop alters a node altitude and all neighbors are updating their tables according to this change.



When a node starts sending a packet, it computes how much time is needed for the next node to send it again as shown in Figure 4. The time delay is the amount of time between sending the packet and receiving a copy when transmitted by next node. Whenever this time is available, upon successful reception of the transmitted packet using promiscuous mode, the node updates the average time delay value. It keeps tracking of this by computing the running average of the time needed by its neighbor to forward its packets. This value is stored in altitude table.

$T_{k}(t) = \gamma T_{k}(t-1) + (1-\gamma)C_{k}(t), \quad \gamma \in [0,1]$ (8)

where $C_k(t)$ is new delay in node *k* at time *t*, T_k is time delay for node *k*. *T* initially is set to be equal to the time required by an unloaded node to send a packet. γ is a parameter regulating how quickly the formula adapts to new information (set to 0.7). Using this type of distance will help in selecting uncongested node. When a node becomes more congested, its time delay will increase. This makes it undesirable and nodes will forward packets to other neighbor. This reflects how water drops behave in nature. When a group of

rivers pour in the same valley, and the amount of outgoing water from the valley is less than the amount of incoming water, a water lake is created. As the water level is increased, the water on some edges will start to leak out to the other sides of the mountains around the lake. Water drops that fall on places close to these edges will follow to other sides with the leaking water rather than to the valley.

When drops move in the network they erode the altitudes of the nodes. The amount of erosion is proportional to altitude difference between sender and receiver nodes. The selected forward node altitude is taken from the altitude table and used to calculate the gradient. The erosion is calculated using

 $erosion_{d}(f) = \propto (altitude_{d}(f) - altitude_{d}(k))$ $altitude_{d}(f) = altitude_{d}(f) - erosion_{d}(f)$

where $erosion_d(j)$ is the amount of erosion at the node *j* toward destination *d*, $altitude_d(j)$ is the altitude of node itself, *k* is the next selected node. α is a positive constant number between 0 and 1 which reflects erosion factor (set to 0.7). High value of α will lead higher erosion and missing of optimal solution. While very low value my lead the algorithm to fall in local minimum. Due to nodes mobility and dynamic infrastructure of mobile ad hoc network, low values for α are not preferred.

At the same time drops add sediment to node altitude. The amount of sediment is proportional to the amount of sediment carried by the drop as well as inversely proportional to the altitude difference of the node and next node altitudes. When the altitude difference is low which represent flat surface, the drop will deposit more sediment.

 $sediment_{d}(f) = (\beta + s * (1 - (altitude_{d}(f) - altitude_{d}(k)))) * carried_sediment, \quad \beta + s \in [0, 1]$ (11)

$$altitude_{d}(f) = altitude_{d}(f) + sediment_{d}(f)$$
(12)

 β and ε are constants controlling the amount of sediments deposit (set to 0.1 and 0.1 respectively). Carried sediment reflects the path characteristic. Paths with higher slops will result in more carried sediment. Setting β and ε to high value will lead to quickly depositing sediments. This will lead to a convex topographical structure where the altitudes around the source are less curvature. While lower values will lead to concave structure and the flat part will close to the destination. Lower value will make drops search around the destination for better delivery nodes. Flat surface around the source will increase the probability of sending data packet to directions opposite to the destination direction, which in turn will move them far away from the destination.

The sediment carried by the drop to next node is

$$carried_{sediment} = carried_{sediment} + erosion_d(f) - sediment_d(f)$$
(13)

Another type of sediment adding occur periodically every fixed amount of time

 $altitude_{dew}(j) = altitude_{dew}(j) + \theta_i \qquad \theta \in [0, 1]$ (14)

where *w* is a set of all known destinations. θ is the amount of sediment to be add(set to 0.01). θ acts as forgetting factor. High value will quickly erase learned information, especially for recent learned paths. Low value is chosen because we don't want to corrupt recent learned information and there are other methods that will also help in adapting the information like the punishment procedure if a link broken. The time period between regular additions is set to 1 second.

To implement smart data protocol, extra fields should be added to data packets in order to act like drops. It is significant to keep the added information to data packets as small as possible and not to overload the data with many extra bytes. Drop packet themselves are small in size. Three parameters have been added to data packet, altitude, carried sediment and the sender address each of them are four bytes in size.

Smart data packets should have the opportunity to discover new routes. This means they should move like drops. In order to limit them from exploring long paths and keep them close to the best known path,

nodes that have minimum altitude and less congested are having minimum distance and considered as best shortest path, data packets are routed using greedy and restricted method. Data packets use altitude table and move according to the following random probability selection function below.

$$P_{n}(j,k) = \begin{cases} \frac{decreasingGradient(j,k)^{2}}{\sum_{k \in V(j)} decreasingGradient(j,k)^{2}} & \text{if } k \in V_{n}(j) \\ 0 & \text{if } k \in V_{k}(j) \end{cases}$$
(15)

where ∂ is a constant number greater than one, and is set to 4 to achieve the greedy movement. V(j) is set of neighbor node of node *j*.

Usually the number of data packets is much higher than drop packets. This high number of data packets could ruin the learning process especially as these packets are moving in a greedy way. This reflects flooding in nature. To limit the erosion that is occurring due to this high number of data packet, the percentage of erosion and sedimentation by data packets is reduced.

4.3 Hello messages and information diffusion

Hello messages are used to propagate information around the network. Essentially, hello message is used to declare node presence; moreover it is used to carry information about node neighbors. Hello message is extended to carry K elements from its altitude table (K=10 is used). Using this idea the altitudes of the nodes will be distributed around the network.

The receiving node will update the altitude toward this node (my_altitudes) and updates its altitude to those destinations that are carried by the hello messages by a factor proportional to the difference between node altitude and received altitude.

$$altitude_{d} = altitude_{d} - \mu(altitude_{d} - hello_{altitude_{d}}), \quad \mu \in [0, 1]$$
(16)

*hello*_{altitude d} is the altitude of destination d in hello message, μ is the adaptation constant (set to 0.1).

Although distributing information by hello messages help to form paths to destinations but it has some drawbacks. The reliability of this information is not high. First, this information does not address the congestion in the nodes. Second, since hello messages are sent every hello interval period this information may be out of date.

4.4 Link failure

The protocol detects route failure in two ways. The first one is by detecting the missing of hello message from a neighbor for a time period more than *allowed hello loss*. *Allowed hello loss* period is set to be twice as *hello interval* which is inherited from AODV protocol. The second way of detecting link failure is through missing acknowledgment after sending data or drop packet.

If a route failure occurred, the node deactivates the route in route altitude table. Then it updates its destination attitude table. If the loss of a neighbor affects the table then it will broadcast a notification to its neighbors.

Whenever the reason of link failure is due to the failed transmission of data packet, than the node will try to send the packet to the next best neighbor. At the same time if the altitude of this neighbor is higher than the altitude of the node itself, it will change its altitude to the same amount of the next best neighbor. In a worse case, if there is no node to deliver the packet to, the node will return the packet to its sender. In this case the altitude of the node to this destination becomes one. The previous node and all neighbors will update their tables with the new altitude value of that node and will route further new packets according to new altitudes. When a node becomes a dead end, there is an opportunity that this node and its neighbors are constructing a local valley. Local valleys should be filled quickly as they work as attracting zones for packets which in turn increases the number of lost packets. The above procedure will prevent wasting many packets on local valleys. Without this procedure, the number of packet required to fill a valley is at least equal to the number of nodes between the dead end node and a node with valid link to the destination. To

prevent wasting this amount of packets, this procedure will damp it with the same packet that is returned. At the same time, the packet is not lost and the learning process is going on.

4.5 Smart data packets routing example

The section explains how smart data packets are routed in the network. In Figure 5.a, node 1 sends packets to node 9. The numbers over each node represent nodes altitude toward node 9. Red arrows represent best route. To explain different approaches of how the routing problem is solved and the advantages of the proposed protocol, we will assume all protocols start by selecting the route 1-2-4-7-9. After selecting the route, we assume that node 4 has involved in a communication session with node 10. Then node 4 moves far from node 2 and the link is broken between them. Node 4 becomes congested and the route is not available as shown in Figure 5.b.

Starting with AODV protocol, when node 4 becomes congested, AODV has no mechanism to detect congestion and it waits until route break occurs. After all, a new route set up procedure is required.

If an ACO based protocol like AntHocNet is used; first let us assume that the rate of ant generation is equal to one ant per second. When route break occur and as AntHocNet is multipath protocol, node 2 forwards the packet to next available node (node 5) and the route become 1-2-5-7-9. Both nodes 5 and 7 are in the transmission of node 4 and will be affected by it, however the protocol does entirely depends on ant agents to detect network status. The protocol will wait for at least one second, if not more, until an ant pass through the route and change the pheromone intensities of the links in that route (by the backward ant).

In the proposed protocol, as node four becomes congested, its distance become higher (equations 5 and 8). The time needed to send the packet from node 2 to node 4 will increase. Node 1 will update the time of node 2 according to equation 8.

Accordingly, in node 1, the probability of selecting node 2 or 3 changes (probability of selecting node 2 decreased depending on the delay). When the link breaks, node two will forward the data to node 5 at same time it changes its altitude to be equal to node 5 (0.45). Node 1 will get a copy of the sent packet and

update the altitude of node 2 in its altitude table to 0.45. Now the altitude of node 3 is less than node 2, therefor node 3 will have higher probability to be selected as next node and the route will change to 1-3-5-7-9. The data packets behave in smart way as they always try to avoid congested area in the network. An important point here is all the above occurred during data packets sending. There is no need to wait for one second like ACO based protocol to send an agent and detect the change in the network parameters. The altitudes have been changed with data packets. No need for feedback agents. Moreover, the changes not only affect the node but it will propagate backward and affect previous nodes' altitudes. It should be noted that the process here is not only message forwarding, it is a distributed learning and optimization technique which continuously learn and react according to network status to find best route. Another advantage of using smart packets is it will minimize convergence time. For example, if the number of ants required in order to find a solution (best route) was 100 ants, with a rate of one ant per second, this will take 100 second. As for SMART protocol with the same rate of agents (drops) and with data rate of five packets per second, the time will be shorter. Each second there will be one drop and five smart data packets that act like drops, which means there will be six drops per second. The convergence time will be about 16.66 second.



Figure 5. An example of smart data packet routing.

5. IMPLEMENTATION AND SIMULATION RESULTS

The performance of the proposed protocol was evaluated by comparing it with AntHocNet (swarm based multipath routing protocol) and standard AODV protocols (with local repair) [9]. AODV is chosen because it is a well-known and almost considered as reference protocol in this research area. AntHocNet protocol is a swarm based algorithm and is chosen because it outperforms many ant routing protocols in many aspects [38, 39].

Simulation results are generated using OMNet++ as simulation software. A model for AntHocNet protocol was implemented based on [25]. As for the AODV protocol, the INETMANET add-on package of the OMNet++ is used. An important point worth mentioning here is we tried to match the setting of all protocols as possible. We have set the rate of ant's generation equal to the rate of drop's generation. We equaled the rate of hello messages for all protocol as well. As we used the standard AODV with no modification as a comparison routing protocol, we kept with its setting and any other setting required for proposed protocol was inherited form AODV protocol.

5.1 Simulation environments

In our experiment, three scenarios have been implemented to test our protocol. Previous studies show that ad hoc network can produce best performance if the number of neighbors is between six to eight [21]. However, we have chosen node density close to 6.25 in order to have good connectivity. With this node density, there is a good probability for a node to have multipath to its destination. At the same time, as the environment become more aggressive, nodes speed increases, the probability of link failure increases, which provide a good environment to test our protocol.

In the first scenario, 32 nodes have been randomly placed in a 600 $*600 \text{ m}^2$ environment. Simulation time was set to 200 seconds. The simulations are repeated for twenty times with different seeds.

The medium access control protocol is the IEEE 802.11 DCF. Packet size is 512 bytes. Five mobile nodes selected randomly to act as sources and five other nodes acts as receivers. Each node generates a

packet every 0.2 second. As described in [18], with this size of network (medium size) and low data rate, the probability of route failure increases which offers more aggressive environment to test the protocol. The network remains silent in the first second. The nodes start sending data at third second and keep sending until the end of simulation, which gives one second for some hello packet to be generated before starting data session. Data traffic is generated using constant bit rate (CBR) UDP traffic sources. Two mobility models are used, random waypoint (RWP) model and Gauss Markov (GM) model [40], to test the performance of the protocols.

In the second scenario, the number of nodes increased to 64 and to keep node density almost equal to the first test, the area is increased to 850*850 m². This scenario is used to test the scalability of the protocol, keeping the same node density and increasing network size. Simulation time was set to 400 seconds. The simulations are repeated for twenty times with different seeds. The number of sources is increased to ten as well. Only random waypoint (RWP) model has been used.

In order to test the performance of the protocol under different node densities, a third scenario is introduced. This scenario is similar to the first one, except that we fixed the speed of the nodes to 10 m/s. The number of nodes varied from 20 up to 100 nodes. Random waypoint (RWP) model is used in this scenario. Moreover, to analyze the impact of the traffic load on the performance of the protocol, we use this scenario to test the protocols under different traffic loads. We set the number of nodes to 32, and increase the load by increasing the number of packets sent per second (packets rate).

Other common parameters for all scenarios are set as following. Each node has a radio propagation range of 150m and channel capacity of 54 Mb/s. Hello time intervals is set to 0.5 second. Pause time is set randomly between 0.1 and 1 second.

The following end to end network characteristic has been studied [25, 41].

1-Throughput: is the measure of the total number of successful delivered data bits over simulations time for a specific node, averaged over the number of source-destination pairs.

2-End to end delay: is the measure of average delay of data packets. This is the time from sending the packet from the application layer at the source node to the time that the packet arrives to the application layer at the destination node, averaged over the number of source-destination pairs.

3-Jitter: is the variation of packet delay which is averaged over the number of source-destination pairs, averaged over the number of source-destination pairs.

4-Routing overhead is the total number of control packets sent divided by the number of data packets delivered successfully.

5- Number of route requests: is the total number of route request generated by specific node (RREQ in AODV, forward ant broadcast in AntHocNet, and ant-drop broadcast in SMART), averaged over the number of source-destination pairs.

6- Average number of buffered packets per node: is the total number of packets that have been buffered in MAC layer averaged over number of nodes.

5.2 Results

Figure 6 and 7 show the throughputs of SMART routing protocol compared with AODV and AntHocNet protocols under different nodes speed. It is clearly seen that SMART protocol performs better than both AODV and AntHocNet. Figure 6 shows that a significant increase in throughput can be achieved by using SMART protocol. For 10 m/s the throughput increased up to 17% over AODV and 13% over AntHocNet under RWP mobility model. When nodes become more mobile, the altitude table, which reflects the topology, will need more updates, therefor the throughput decreases. Comparing the results of the network under RWP mobility with GM mobility, we can see the protocol preforms better under RWP mobility. There are two main reasons for that. Firstly, in RWP mobility, the nodes tend to move toward the center of the simulation area and move away from the simulation area boundary. This will leads to fluctuation in node density and as a result, the path lengths become shorter. The nodes are better distributed

under GM mobility model. Secondly, nodes movement in GM mobility is correlated. In RWP, node can make a sudden change in its direction independent on its previous direction. These two reasons led to better results under RWP mobility model.

Figure 7 shows the throughputs for 64 node network under RWP mobility. The throughput increased about 30% over AODV in low speed and about 13% over AntHocNet. We can notes that SMART protocol preforms better than the other protocols because it uses data packet in learning process. Comparing Figure 6 and 7, the effect of network size and number of sources is obvious and the throughput for all protocol decreases.



Figure 6. Average throughput for 32 nodes network under various speed values.



Figure 7. Average throughput for 64 nodes network under various speed values

The end to end delay for the first scenario is shown in Figure 8 and for the second scenario is shown in Figure 9. Both figures show more than 96% enhancement in end to end delay. SMART data protocol produces better results than the others. The main factors that contribute in network latency and jitter are congestion, queuing and route changing. In AODV most of the delay occurs due to route setup and broadcasting of control packets and route maintenance as well as queuing of data packets. At the same time, the criteria of selecting next forwarding node and finding the optimal path from source to destination also contributes in the delay. SMART protocol minimizes the number of broadcasting, and tries to select a non-congested node. If a node had a route failure, it forwards the packet to other best node. At the same time when a packet is moving around the network, the process of learning is still carried on and nodes update their altitude tables continually.

AntHocNet forward data depending only on regular pheromone (there are two type of pheromones tables in AntHocNet, regular and virtual) which mostly trained by backward ants. This table may not have all possible paths to destinations as this require huge number of ants and training cycles, so in many situation route break occurs. The method of packet forwarding and continues real time learning of SMART protocol leads to better performance. Smart data packets allow the network to learn more rapidly and the packets can

find other routes. Another important factor, if a better route is found, the AntHocNet needs many ants to enhance the weight of the route to be selected. SMART protocol is based on RFD and adopts faster.

Comparing Figure 8 with Figure 9, the effect of network size and number of sources is obvious. As the network become larger and number of source nodes increases, the probability of link breakage increases and the repair time will be longer as well as the network becomes more congested. As the network size increases, the search space for finding better routes also increases.



Figure 8. Average end to end delay for 32 nodes network under various speed values.







Figure 10. SMART protocol end to end delay under various speed values.

The end to end delay of the SMART protocol is very low. Figure 10 shows the effect of speed on the end to end delay for the first and second scenarios. For 32 node network, the average end to end delay under both RWP and GM mobility is between 0.89 up to 1.3 millisecond and increasing slowly as node speed increases. The delay is higher in 64 node network and starts from 1.4 millisecond up to 1.8 millisecond.

According to network setting, the time needed T_p to send a packet in our network as we use two way handshaking (DATA- ACK) could be calculated as below [42-44]

$T_{p} = T_{DIFS} + T_{SIFS} + T_{BO} + T_{DATA} + T_{ACK}$ (17)

where T_{DIFS} is the Distributed InterFrame Space time (34 µs), T_{SIFS} is Short InterFrame Space (SISF) time (9 µs), T_{BO} is backoff interval time (min 67.5 µs), T_{DATA} is MAC Protocol Data Unit (MAC PDU) plus Physical Layer Convergence Protocol(PLCP) header plus PLCP preamble (for 512 byte plus extra fields, it will be 106.6 µs), and T_{ACK} is acknowledgement time (24 µs). For 32 node network, in average, the end to end delay time is more than four successful data transmission time, while it is equal to about seven successful transmissions in 64 nodes network.

The extra bytes that been added to smart data packet will add about 2 microsecond to the transmission time of data packet, which is less than 1% of data transmission duration. If a separate control packet has been used for this information it would introduced a lot of delay to the network. Sending 12 bytes as separate packet including MAC control overhead will require 161 microsecond. Apart from the total number of control packet required if a separate control packet is used, the reduction in time is obvious. As an example of reduction of control packets, the average throughput at 10m/s for 32nodes network is 18281bps, the average delivered packet to each destination is 883. If separate packet used, this would require extra 883 control packets to be delivered to each destination. In spite the extra time needed to transmit a smart data packet, its efficiency is cleared from the above.

Figure 11 and 12 show the jitter for the first and second scenarios respectively. Again SMART protocol overcome both protocols and has less jitter. Variation in network structure due to nodes mobility causes link breakage. Both AODV and AntHocNet have route recovery mechanism which consists of queuing and rebroadcasting. SMART protocol is based on instantaneous data packet rerouting to other better route in case of route failure. As the data packets moves from source to destination, all nodes within the route will continue their learning process by the data packet. Moreover, all nodes that are neighbors to the route will also learn as the process of learning in SMART protocol is distributed. An extra point to address here, data packets are routing themselves through multipath to the destination through nodes around best discovered path. This will load balance the network, as well as it creates a valley like structure which its lowest end is at the destination. When a node in-between becomes unreachable, the data packets should easily find another path to the destination.



Figure 11. Average jitter for 32 nodes network under various speed values.



Figure 12. Average jitter for 64 nodes network under various speed values.

Both AntHocNet and SMART protocols are hybrid routing protocols and they send control messages throughout the entire data session in order to maintain the route between the source and the destination. This could produce more overhead in the network and costing the network to use more resources. Figure 13 shows the control packet overhead of the three protocols for both the first and second scenarios under RWP mobility. SMART protocol generates more control overhead than AODV protocol, however the control overhead is less than AntHocNet. The absence of backward agent and the use of SMART data packets led to less control overhead than AntHocNet.



Figure 13. Control packet overhead under various speed values.

The third scenario is used to study the effect of node density on the protocol performance. Figure 14 shows the effect of node density on network throughput. When the number of nodes is less than 40 nodes in the network, the connectivity of the network is degraded and the throughput decreased. When the number increases over 60 nodes, the network become more congested and the throughout decreases again. From the figure it is clear that the proposed protocol overcome both other protocol throughout all number of nodes.

Figure 15 shows the end to end delay for various numbers of nodes. When there are few nodes in the network, the number of route failure increases. For each route failure, the AODV route recovery requires more time which causes more delay because of broadcasting. AntHocNet has faster recovery procedure [7], however it also uses broadcasting algorithm whenever a route break occurs. SMART protocol usually do not buffer data packet, unless if the destination is unreachable. As the network become denser, the number of nodes involved in a route may increase in both AODV and AntHocNet. The cost of broadcasting increases as well as the probability of collision increases. SMART protocol is more efficient in searching for better routes, and it minimizes the number of broadcasting in the network. It can be seen that the end to

end delay at low node density is high. This occurs because at low node density the probability of the destination being unreachable is high. The delay occurs usually as a consequence of queuing and broadcasting. Putting in mind, in our protocol queuing occurs only at the original source when the destination is unreachable and all the nodes around the original source do not have a route to the destination. In other words all the nodes around the source have an altitude equal to one. This situation is rarely occurs unless the source node is isolated alone or with only few nodes. The reason that reduces this situation is hello message and distributed learning. Whenever a drop is moving through the network it erodes its path, and node surrounding the path will be also eroded by distributed learning. Hello message is also eroding the altitudes of neighbor nodes. However, two methods are participating in increasing the altitudes, the sediment addition, and the punishment process. The ratio of sediments addition is always low. The punishment in our approach is limited to two nodes to decrease network traffic and prevent the losing of learned information. Limiting the punishment procedure to two nodes, described in section 4.4, will require many packets to set the altitude of all the nodes around the source to one. Moreover, as explained in section 4.3 that hello messages erode altitude of neighbor nodes, if a node with an altitude less than one broadcasts its hello messages to those neighbor that have been punished, it will result in decreasing their altitudes. This will also decrease the probability of bringing the altitudes of all nodes around the source to one. Another factor also contribute in decreasing this probability is when a node joins these group, and it has an altitude lower than one to the desired destination, it will decrease the altitude of its neighbor in the group when it send hello messages or when a data activity occur at that node as all the neighbors will listen to its activity (promiscuous mode updating). Smart data packets as well as drops may be forwarded to that node. If this node joins another group to the original source group and become a link between them, smart data packets and drops will try to find a route to destination through this new group, and in case if there is a valid path to the destination from that node, they will follow that route. Even if the topology of these nodes and links changed and become expire, these smart packets will adapt and search for a link to the destination. After all, this is why we can see that the end to end delay in our protocol is always low, as data

packets are always being sent and searching and creating new routes. The probability of queuing as well as broadcasting is low and data packets may be deleted in case if they reach a dead end or when maximum number of hops is reached. Throughout all our simulation, we observed that the maximum number of broadcasting for any source (route set up) in each simulation was always less or equal to three. This indicates that the probability of queuing is very low. Table 3 shows the average number of route request generated by source nodes in AODV protocol compared to SMART protocol. AODV has a local route repair procedure and the average number of route repair at each node in the network is also shown in the table. The results are for scenario one and for random way point mobility. The effectiveness of the RFD algorithm can be seen as drops and smart packets are always moving and searching for a route rather than depending on a recovery mechanism as in AODV or AntHocNet. Whenever a link becomes expire, drops as well as smart packets will follow to new offered paths to search for destination. Similarly, when a flow of water drop is closed, water drops will follow new paths until they reach the sea. In their way to the sea, they will continue their erosion and sedimentation process, i.e. the learning process is never stops.



Figure 14. Network throughput under various numbers of nodes.



Figure 15. Network end to end delay under various numbers of nodes.

	Speed (m/s)					
	10	20	30	40	50	
Number of Route discovery for AODV protocol	19.95	26.45	32.24	35.81	39.28	
Local route repair for AODV protocol	2.18	2.62	2.48	2.59	2.70	
Number of Route discovery for SMART Protocol	1.22	1.24	1.34	1.38	1.46	
Number of Route discovery for AntHocNet	12.54	16.38	22.02	26.66	28.56	

Table 3: Average number of route request for the first scenario under RWP mobility

Figure 16 shows the effect of increasing traffic load on the proposed protocol compared to AODV and AntHocNet protocols. The number of packet per second varied from 10 packets to 60 packets per second. It is clear that the throughput of SMART protocol is better than others especially in high traffic load at 60 packets per second. Figure 17 shows the end to end delay under variable packet rate. The negative slope of the AODV end to end delay at the beginning is due to the rate of encountering route failure for low data rate is high or the probability of finding difficulties to build valid route to the destination [18].



Figure 16. Network throughput under various traffic load.



Figure 17. Network end to end delay under various traffic load..



Figure 18. Low packet data rate example

Table 4 shows the average number of packets buffered at MAC layer per second. It is clear that the main cause of delay at high data rate is because of buffering at MAC layer. The design of MAC protocol and the detection of link failure at MAC layer caused more packets to be buffered and added delay to the network. At low data rate, the source of delay is due to network layer protocol. The number of packets buffered at MAC layer is very low.

Figure 18 shows an example of the reason that at low data rate the delay is high in AODV protocol and decreasing as the data rate increases until certain data rate. In figure 18a, the ratio of packets that suffering from delay to overall number of packets passing through a node is ½ (1 packet delayed per 2 packets). When the data rate increases, as in figure 18b the ratio will decreases and as example become ¼ so the average delay decreases as data rate increases until certain data rate. It should be noted that when a source node is dealing with route recovery the new packets will be buffered and network layer until a route will be found.

The SMART protocol rarely has route recovery; however the delay of link breakage detection occurs as it is related to MAC and link layers protocols. The link break detection depends on the retry limit constant set at MAC layer which define how many time the node will try to send the packet before reporting link breakage. The delay in AODV will be higher as route recovery based on broadcasting while the SMART protocol will forward the packet to next available node.

At a certain data rate, the traffic in the network will suffer more delay as network become congested and the number of packet that are buffered in both network and MAC layer increases resulting in more end to end delays. As this delay at high data rate is inherited from MAC layer protocol, we preferred to use low data rate to show the delay caused by network protocol.

data rate						V				
(p/s)	10	20	30	40	50	60	70	80	90	100
AODV	0.105469	0.196387	0.394531	14.28906	39.97314	246.9009	886.0752	1209.461	1344.716	1791.581

25.60547

170.8007

533.3301

743.4785

1274.656

1622.425

14.35547

0.098633

SMART

0.188477

0.323242

Table 4: AVERAGE NUMBER OF BUFFERED PACKETS PER NODE

6. CONCLUSIONS

In this paper we have proposed smart data routing protocol for mobile ad hoc networks based on the RFD algorithm. RFD is a swarm algorithm inspired by the way rain drops make rivers.

The learning in the RFD algorithm is feed forward and eliminates the need for backward packets. This reduces the number of control packets in the network and offers a good opportunity to change and implement smart data packets.

Data packet in the proposed protocol could be ordinary or smart packet. The proposed protocol is flexible and can work on both smart and ordinary data packet. Smart packets carry extra fields in order to contribute in learning process which as result affects the movement of the packets in the network. These extra fields

are appended to the end of IP packet header, which adds more compatibility and flexibility to the protocol in order to handle ordinary data packets.

Our results show that smart data protocol performs better than AODV and AntHocNet. In average, the throughput is increased and both end to end delay and jitter decreased.

References

[1] P. Lalbakhsh, B. Zaeri, A. Lalbakhsh and M.N. Fesharaki, "AntNet with Reward-Penalty Reinforcement Learning," *Computational Intelligence, Communication Systems and Networks (CICSyN), 2010 Second International Conference on*, 2010, pp. 17-21.

[2] A. Boukerche, B. Turgut, N. Aydin, M.Z. Ahmad, L. Bölöni and D. Turgut, "Routing protocols in ad hoc networks: A survey," *Computer Networks*, vol. 55, no. 13, 2011, pp. 3032-3080.

[3] Kwang Mong Sim and Weng Hong Sun, "Ant colony optimization for routing and load-balancing: survey and new directions," *Systems, Man and Cybernetics, Part A: Systems and Humans, IEEE Transactions on*, vol. 33, no. 5, 2003, pp. 560-572.

[4] S. Marwaha, J. Indulska and M. Portmann, "Biologically Inspired Ant-Based Routing in Mobile Ad hoc Networks (MANET): A Survey," *Ubiquitous, Autonomic and Trusted Computing, 2009. UIC-ATC '09. Symposia and Workshops on,* 2009, pp. 12-15.

[5] G.S. Sharvani, N.K. Cauvery and T.M. Rangaswamy, "Different Types of Swarm Intelligence Algorithm for Routing," *Advances in Recent Technologies in Communication and Computing*, 2009. ARTCom '09. International Conference on, 2009, pp. 604-609.

[6] B. Kalaavathi, S. Madhavi, S. Vijayaragavan and K. Duraiswamy, "Review of ant based routing protocols for MANET," *Computing, Communication and Networking, 2008. ICCCn 2008. International Conference on, 2008, pp. 1-9.*

[7] P. Jacquet, P. Muhlethaler, T. Clausen, A. Laouiti, A. Qayyum and L. Viennot, "Optimized link state routing protocol for ad hoc networks," *Multi Topic Conference*, 2001. *IEEE INMIC 2001. Technology for the 21st Century. Proceedings. IEEE International*, 2001, pp. 62-68.

[8] C.E. Perkins, P. Bhagwat, C.E. Perkins and P. Bhagwat, "Highly dynamic Destination-Sequenced Distance-Vector routing (DSDV) for mobile computers;" *SIGCOMM Comput.Commun.Rev.*, vol. 24, no. 4, 1994, pp. 234-244.

[9] C.E. Perkins and E.M. Royer, "Ad-hoc on-demand distance vector routing," *Mobile Computing Systems and Applications, 1999. Proceedings. WMCSA '99. Second IEEE Workshop on*, 1999, pp. 90-100.

[10] D.B. Johnson, D.A. Maltz and J. Broch, "DSR: The Dynamic Source Routing Protocol for Multi-Hop Wireless Ad Hoc Networks," *Addison-Wesley*, 2001, pp. 139-172.

[11] P. Samar, M.R. Pearlman and Z.J. Haas, "Independent zone routing: an adaptive hybrid routing framework for ad hoc wireless networks," *IEEE/ACM Trans.Netw.*, vol. 12, no. 4, 2004, pp. 595-608.

[12] L. Guo, L. Zhang, Y. Peng, J. Wu, X. Zhang, W. Hou and J. Zhao, "Multi-path routing in Spatial Wireless Ad Hoc networks," *Computers & Electrical Engineering*, vol. 38, no. 3, 2012, pp. 473-491.

[13] E. Gelenbe, Zhiguang Xu and E. Seref, "Cognitive packet networks," *Tools with Artificial Intelligence, 1999. Proceedings. 11th IEEE International Conference on*, 1999, pp. 47-54.

[14] Z. Wang and J. Crowcroft, "Quality-of-service routing for supporting multimedia applications," *Selected Areas in Communications, IEEE Journal on*, vol. 14, no. 7, 1996, pp. 1228-1234.

[15] P. Rabanal and I. Rodriguez, "Hybridizing River Formation Dynamics and Ant Colony Optimization," in Advances in Artificial Life. Darwin Meets von Neumann, vol. 5778. George Kampis, et al., Springer Berlin Heidelberg, 2011, pp.424-431.

[16] P. Rabanal, I. Rodríguez and F. Rubio, "Using River Formation Dynamics to Design Heuristic Algorithms," *Unconventional Computation Lecture Notes in Computer Science*, vol. 4618, 2007, pp. 163-177.

[17] M.K. Marina and S.R. Das, "On-demand multipath distance vector routing in ad hoc networks," *Network Protocols*, 2001. *Ninth International Conference on*, 2001, pp. 14-23.

[18] B. Blanco, F. Liberal and I. Taboada, "Suitability of ad hoc routing in WNR: Performance evaluation and case studies," *Ad Hoc Networks*, vol. 11, no. 3, 2013, pp. 1165-1177.

[19] V.R. Budyal and S.S. Manvi, "ANFIS and agent based bandwidth and delay aware anycast routing in mobile ad hoc networks," *Journal of Network and Computer Applications*, no. 0.

[20] G. Singh, N. Kumar and A. Kumar Verma, "Ant colony algorithms in MANETs: A review," *Journal of Network and Computer Applications*, vol. 35, no. 6, 2012, pp. 1964-1972.

[21] R. Schoonderwoerd, J.L. Bruten, O.E. Holland and L.J.M. Rothkrantz, "Ant-based load balancing in telecommunications networks," *Adapt.Behav.*, vol. 5, no. 2, 1996, pp. 169-207.

[22] G. Di Caro and M. Dorigo, "Mobile agents for adaptive routing," *System Sciences, 1998., Proceedings of the Thirty-First Hawaii International Conference on*, vol. 7, 1998, pp. 74-83.

[23] S. Marwaha, Chen Khong Tham and D. Srinivasan, "Mobile agents based routing protocol for mobile ad hoc networks," *Global Telecommunications Conference*, 2002. *GLOBECOM* '02. *IEEE*, vol. 1, 2002, pp. 163-167.

[24] M. Gunes, U. Sorges and I. Bouazizi, "ARA-the ant-colony based routing algorithm for MANETs," *Parallel Processing Workshops, 2002. Proceedings. International Conference on,* 2002, pp. 79-85.

[25] G.D. Caro, F. Ducatelle and L.M. Gambardella, "AntHocNet: An adaptive nature-inspired algorithm for routing in mobile ad hoc networks," *European Transactions on Telecommunications*, vol. 16, SEP 2005, pp. 443-455.

[26] M. Rafsanjani, S. Asadinia and F. Pakzad, "A Hybrid Routing Algorithm Based on Ant Colony and ZHLS Routing Protocol for MANET,", vol. 120, 2010, pp. 112-122.

[27] H.F. Wedde, M. Farooq, T. Pannenbaecker, B. Vogel, C. Mueller, J. Meth and R. Jeruschkat, "BeeAdHoc: an energy efficient routing algorithm for mobile ad hoc networks inspired by bee behavior," *Proceedings of the 2005 conference on Genetic and evolutionary computation*, 2005, pp. 153-160.

[28] E. Gelenbe and R. Lent, "Power-aware ad hoc cognitive packet networks," *Ad Hoc Networks*, vol. 2, no. 3, 2004, pp. 205-216.

[29] R. Lent, "Smart packet-based selection of reliable paths in ad hoc networks," *Design of Reliable Communication Networks*, 2005. (DRCN 2005). Proceedings.5th International Workshop on, 2005, pp. 5 pp.

[30] L. CHEN, H. JI, Y. LI and X. LI, "Multi-path routing based on load-balance for cognitive packet networks," *The Journal of China Universities of Posts and Telecommunications*, vol. 18, no. 5, 2011, pp. 71-75.

[31] Hifza Afaq, Sanjay Saini, "On the Solutions to the Travelling Salesman Problem usingNature Inspired Computing Techniques," *IJCSI International Journal of Computer Science Issues*, vol. 8, no. 4, July 2011, pp. 326 -334.

[32] P. Rabanal, I. Rodríguez and F. Rubio, "Applying RFD to Construct Optimal Quality-Investment Trees," *J.UCS Journal of Universal Computer Science*, vol. 16, no. 14, May 2010, pp. 1882-1901.

[33] P. Rabanal, I. Rodríguez and F. Rubio, "Applying River Formation Dynamics to the Steiner Tree Problem," *Cognitive Informatics (ICCI), 2010 9th IEEE International Conference on,* 2010, pp. 704-711.

[34] P. Rabanal, I. Rodriguez and F. Rubio, "A Formal Approach to Heuristically Test Restorable Systems," in Theoretical Aspects of Computing - ICTAC 2009, vol. 5684. Martin Leucker and Carroll Morgan., Springer Berlin Heidelberg, 2009, pp.292-306.

[35] P. Rabanal, I. Rodriguez and F. Rubio, "Finding Minimum Spanning/Distances Trees by Using River Formation Dynamics," in Ant Colony Optimization and Swarm Intelligence, vol. 5217. Marco Dorigo, et al., Springer Berlin Heidelberg, 2008, pp.60-71.

[36] P. Rabanal and I. Rodriguez, "Testing Restorable Systems by Using RFD,"in Bio-Inspired Systems: Computational and Ambient Intelligence, vol. 5517. Joan Cabestany, et al., Springer Berlin Heidelberg, 2009, pp.351-358.

[37] M.K. Marina and S.R. Das, "On-demand multipath distance vector routing in ad hoc networks," *Network Protocols*, 2001. *Ninth International Conference on*, 2001, pp. 14-23.

[38] B.B. Vasundhara Uchhula, "Article: Comparison of different Ant Colony Based Routing Algorithms," *IJCA Special Issue on MANETs*, no. 2, 2010, pp. 97-101.

[39] J.S. E, "Performance Comparison of ACO Algorithms for MANETs," *International Journal of Advanced Research in Computer Engineering & Technology(IJARCET)*, vol. 2, no. 1, 2013.

[40] C. Bettstetter, "Smooth is better than sharp: a random mobility model for simulation of wireless networks," *Proceedings of the 4th ACM international workshop on Modeling, analysis and simulation of wireless and mobile systems, 2001, pp. 19-27.*

[41] S. Rajagopalan and C. Shen, "ANSI: a swarm intelligence-based unicast routing protocol for hybrid ad hoc networks," *Journal of Systems Architecture: the EUROMICRO Journal*, vol. 52, no. 8, 2006, pp. 485-504.

[42] Jangeun Jun, P. Peddabachagari and M. Sichitiu, "Theoretical maximum throughput of IEEE 802.11 and its applications," *Network Computing and Applications, 2003. NCA 2003. Second IEEE International Symposium on, 2003, pp. 249-256.*

[43] Shao-Cheng Wang, Y.-. Chen, Tsern-Huei Lee and A. Helmy, "Performance evaluations for hybrid IEEE 802.11b and 802.11g wireless networks," *Performance, Computing, and Communications Conference, 2005. IPCCC 2005. 24th IEEE International*, 2005, pp. 111-118.

[44] P. Raptis, V. Vitsas and K. Paparrizos, "Packet Delay Metrics for IEEE 802.11 Distributed Coordination Function," *Mob.Netw.Appl.*, vol. 14, no. 6, 2009, pp. 772-781.

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