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Two energy efficient algorithms for tracking objects in a sensor network

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Summary

We propose two energy efficient algorithms for locating a target object moving in an area covered by a wireless ad hoc network. The first algorithm developed conserve energy by efficiently identifying sensor nodes, as *Home Nodes*, and use only local messages between neighboring nodes to follow the trail of the object. Since we avoid the long-range transmission and maximize the localization, the algorithms reduce the communication cost. The dynamic nature of the second algorithm exploits the predefined parameters such as the object velocity. Our algorithm represents query shipping against the conventional data shipping as a means to reduce the amount of data being shipped across the network. Hence, it locates the objects over the network with minimal energy conservation using short-range message transmissions. The performance analysis (both experimental and theoretical) shows the effectiveness of the two algorithms in comparison to another tracking algorithm. Copyright © 2006 John Wiley & Sons, Ltd.

KEY WORDS: wireless ad hoc network; sensor nodes; energy efficient; algorithm, communication

1. Introduction

A wireless sensor network is comprised of compact, autonomous nodes equipped with data gathering, computation, and wireless communication capabilities. Wireless sensor networks [1,8,9,16] are considered to have the potential to be extremely useful in emergency situation where little or no infrastructure is available. Tracking moving objects is an interesting problem with many applications. Such a network can be used to track children in a hospital and market place. Think of a toddler wanders off without your knowledge. With a location tracking system in place, one can easily track kids in a large crowded area. Through the use of a concealed wireless tracking

device, such systems can aid tremendously if someone tries to take a child without permission. Also, the use of location-aware applications in real environments is very exciting [10,11,13]. One can deploy this type of system in a shopping mall or large retail store to broadcast electronic flyers and advertisements. The system takes into consideration the physical location of shoppers within the facility and customizes the advertisement appropriately. For example, users is receive an electronic directory and advertisement flyer on their PDA after entering the mall. The directory would include a map of the facility that identifies the person's exact position. As the shopper clicks on a store, restroom, or ATM machine in the directory, the map indicates directions that take them to the desired

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selection. Furthermore, retailers can push sales information to shoppers that come within range of the store. If the user selects one of the advertisements, their PDA can lead them to the applicable merchandise. Of course, this can result in more electronic competition. For example, one is shopping for tools in one store, and a competing store can pop up an advertisement for a better deal in an attempt to steal the business. To use a GPS device [4,17] for such applications, first a line of sight is required. Due to unavailability of GPS signals in building and indoor ambient, tracking an object inside a building is very difficult, and there are sensors may not be equipped with GPS units.

In a highly dynamic and unpredictable ad hoc network, the network topology may change more quickly than it can be tracked. Hence, the control decisions and predictable algorithms based on this domain knowledge will always lag behind the actual state of any mobile device. However, even in cases where real-time tracking of current network state is feasible given the quantity of resources necessary for nodes to keep state information up-to-date should be large enough to preclude the network from performing other functions. Mobile devices may elect not to distribute all state information so that they are able to perform their other necessary functions, and hence mobile devices may have incomplete information about current network state.

In ad hoc networks [14] with resources scarcity, protocols must be designed to consume the minimum amount of computing resources necessary for correct and acceptable operation without performance degradation. Unfortunately, many network protocols today are relatively easy to force into states in which they consume large amount of resources. And, more unhealthy aspect in existing protocols is a malicious node could issue frequent link state changes or could even frequently change the state of one of its links, forcing the routing protocol to generate and distribute many unnecessary updates to the routing information that consumes much of the limited resources. Hence, designing such protocols effectively to manage the resources is a challenging task.

As a wide spread deployment, it is not possible to administer a scalable system when all control functions are centralized. So designing a decentralized system not only provides scalability but it is also easy to install the system on owners' interest. There is no need for a central entity to keep track of object and a decentralized system provides better bandwidth management. One of the most important parameter about

network performance is the bandwidth. There are many issues that can degrade the performance of the entire network and hence, our goal in this project is to develop algorithms that effectively use the bandwidth with minimum level of interference.

The purpose of this research is to track entities which enter the field of vision of nodes in the ad hoc network. Based on objects sightings, nodes can maintain a dynamic cache that can be queried by a base station. A query from the base station will ask for the latest location of any given object to a node in ad hoc Network. Since transmitting messages consumes a lot of energy against local processing, and moreover, the limited energy of nodes in ad hoc networks do not allow heavy duty computing, we present algorithms that will ensure efficient computing and minimal transmission without degradation of performance. Since data is stored locally, and the memory is finite, the nodes must choose their cache lines carefully and may need to dump the redundant data so that it could avoid cache overflow. Our algorithms make use of 'Home Node' concept and 'velocity constraint' to track objects. Our methods minimize the communication among nodes as against the methods proposed [2] where tracking an object may involve traversing the complete mesh. We report the performance evaluation of our algorithms and their comparison.

This paper describes a location tracking system for a moving object. In particular, our two algorithms avoid long-range messages and tracking objects involves entirely local computation within the network. Since structured and small messages are transmitted, we provide an effective memory management at each memory constraint node and thus, increase the life of the node in the network.

2. Related Work

Many papers in literature proposed schemes and methods for localization in ad hoc networks [10,12,15]. Often, triangulation is applied in order to self-estimate the position; in some cases, some special nodes whose position is known called beacons [3] are used for a more accurate estimation. In References [4,6], RF is used to find the location of static nodes inside a building. Brooks *et al.* [5] suggested a cluster-based scheme, where all the cluster nodes exchange sensing data with each other and therefore, is not energy efficient. Cerpa *et al.* [7] suggested multiple nodes to do tracking, however there was no scheme given. In Reference [2], each node in the network does

have a field of vision $F(v)$, which indicates the area where the node detects the object. Once detected, the node will retain the signature of the object and updates the cache based on the cache management protocol. Each ad hoc node is able to retain signatures of all objects detected in its cache, and the cache is effectively unbounded. An object traces a continuous curve as it moves in the area covered by the sensors. The sensors register the object traces instant the object entered the monitored area until it left the area. If the sensors cover an area completely then at any instant, the object was in the field of vision of at least one sensor node. Correspondingly, as the object moved out of the field of vision of a sensor, it moved into the field of vision of a neighboring sensor node [2].

The algorithm [2] finds the most recent location of a given target objects but the reduction in the number of messages depends on the movement patterns of the object. In this algorithm, sending more messages when the object velocity is high will flood the network with short-range messages and therefore, the performance will gradually decrease as the high velocity objects are tracked. Moreover, the time to track objects across the network will increase since the query will be processed over the entire network. The static nature of detection also makes the system less scalable and adaptable. Given a cut, if the target object *ever* passed from one part of the area to the other, at least one node on the cut must have seen the target object. The process of picking up the trail is reduced to picking an appropriate set of cuts, and querying along the cuts for a sighting of the target object [2]. Hence, defining a proper cut is also essential and integral part of the design. A static cut will degrade the performance since the base station will query each leader on misidentification or on detecting a static object.

The object tracking system in Reference [2] uses a *query shipping* architecture. Unlike data shipping, query shipping involves tracking the location of an object through short queries. Queries will be transmitted from an external base station to the node of the cut of sensor network nodes. The nodes coordinate among themselves with short query messages and after detecting the node with latest time stamp, it will inform about the latest location to the base station. However, with the data being stored locally, there is a need of an effective memory management for each node.

The query shipping architecture provides a better service than any conventional message shipping solutions for following reasons: queries are small in size compared to bulk data passing and are, therefore, more effective when tracking a location of an object.

Also, the system will be more fault tolerant compared to data shipping; the sensor node can retrieve information on the loss of message during transmission.

3. Proposed System Model

In this paper, we discuss a resource limitations aware solution to the problem such as memory management and light-weight algorithms for detecting the location of a physically moving target object. Our solution would be to ship the query to a designated node, which later queries the neighbors across the network using hop-by-hop routing. Any sensor node, which has seen the object, responds back to the base station, with its time of sighting. We introduce a concept of 'Home Node' called HN. It is like any other node in the network except that it acts like a home for the object that responds with the latest time stamp to the query from the base station.

We propose a solution that optimizes the computing resources for a given node without performance degradation. Since physical objects are continuous in space and time, if the target object moved from one portion of the area to another, some node on the boundary separating the two portions must have seen the object. Moreover, objects leave a trail of entries in the caches of sensor nodes along their path. Thus, we search for a node, which has observed the target object by taking a sequence of home nodes and their neighbors in the sensor network. Once such a node is found, the trail left by the object is followed to its current position. We propose a class of algorithms that build on the basic ideas above to minimize the number of long as well as short-term message transmissions. Here, we adhere to the naming system of Reference [2]. Also, we propose a new idea of multicasting the short-range messages depending on the velocity of the object. We define a cut based on velocity and then multicast the query across the cut, which will be discussed in detail in later section.

In this paper, we use the same naming convention as Reference [2], and in addition, we introduce some new parameters.

- $V = \{V1, V2, V3 \dots\}$ is a set of sensor nodes in a geographic region R
- Nid is a unique identifier of a sensor node and each node has a field of vision: $F'(v)$
- $Nbrs(v)$ is a set of sensor node's neighbors
- B is the base station which is aware of the topology of the sensor network

- *Lid*: Logical identifier of the object
- C: Cache memory
- Cptr: Cache pointer
- TS: Time stamp
- LS: Last access time stamp
- $T(X, Lid)$ is the time of sighting of the target *Lid* by *X*
- Q: (*Lid*, *X*, $T(X, Lid)$), where *Lid* is the target's id, and *X* will be the sensor sighted the target at time *T*
- HN(*Lid*): home node of the object
- SR: Short range distance

Here, any node in the sensor network can communicate to base station and vice versa. We avoid long-range messages for better resource utilization (power) without performance degradation. Moreover, the local computation within the sensor network with less long-range message will not only increase security but also better bandwidth management. Each node in sensor monitors certain area denoted by $F(v)$; the field of vision. The sensor input captured is converted into a signature *Lid*; a logical identifier. The signatures of objects observed by a node are recorded in its cache, along with the time of its observance.

The base station transmits queries to the sensor network. Unlike Reference [2], the base station queries the sensor node, which had reported the object location on the pervious query. Next, the queried sensor node will in turn query the neighbors about the current location of the object. The queries will be of the form Q: (*Lid*, *X*, $T(X, Lid)$), where *Lid* is given, while *X*, *T* are free. In other words, *Q* asks for the location of a target with signature *Lid*.

The node with the latest time stamp ($T(X, Lid) > T(Y, Lid)$) for all *Y* in the sensor network, reports to the base station. The corresponding node will be queried whenever the base station wants to track the object again. It can probe the network for the location of a specified target. The sensors coordinate with each other to find a suitable answer for the query. Since it is desirable that the sensor network has as long a life as possible, the sensors should transmit few messages as possible. In addition, each sensor should have similar duration of life to prevent 'holes' in coverage.

4. Algorithm for Tracking Objects Using Home Node

Here, we describe algorithm that will track the objects. The base station receives a query *Q* that looks

for an entity with signature *Lid*. The base station chooses a node as the home of the object, and later whenever the base station needs to locate the object the home node will be queried. Here, we assume that each tracking object has a 'Home Node' which has sighted the object last. This list is updated at the base station each time the object is located.

The Figure 1 describes the basic functions of the location tracking system. The base station queries the home node (HN) for tracking the latest location of the object. Later the HN, queries its Nbrs for its latest position. The above process ends when the network finds the latest location of the object. The node will report to the base station and in turn, the base station updates the home node against the *Lid*.

Algorithm Track Object: Find a home node of the object.

Input: *Lid*: Logical identifier of the target object
HN (*Lid*): Home Node of the *Lid*.

Output: *X*: Identifier of a sensor, which saw the target object

Method:

- (1) Query all nodes in Nbrs of HN (*Lid*)
- (2) If no nodes in Nbrs of HN(*Lid*) have seen *Lid* later, send(*Nid*, $T(Nid, Lid)$) to the base station.
- (3) Else each node sends its $T(Nid, Lid)$ if it is greater than $T(HN, Lid)$ else drop the message.
- (4) Query the node with the latest $T(Nid, Lid)$ to do TrackObject;
- (5) Return(*X*);

The 1st phase of the algorithm returns the identifier of the node with the most recent sighting time and is assigned as the home node of the object. The base station will contact the HN to continue to track objects. The elected node, HN, then starts the process of following the trail of the target object. If the target object moved out of range of the sensor, it must have gone into the range of one of its neighbors. By finding the particular neighbor, the elected node who saw the target object last can determine the next step in the trail of the target object. Such a neighbor is then elected as the home node to follow the target object.

The trail stops when a node on the trail had seen the target object after *entire* neighboring area is searched. The trail stop implies that either the target object is still in the range of this existing node, or the elected node is an edge node and the target object has left the grid of sensors. In either case, we found the most recent sighting of the target object. Such a node then responds back to the base station giving its identifier and the

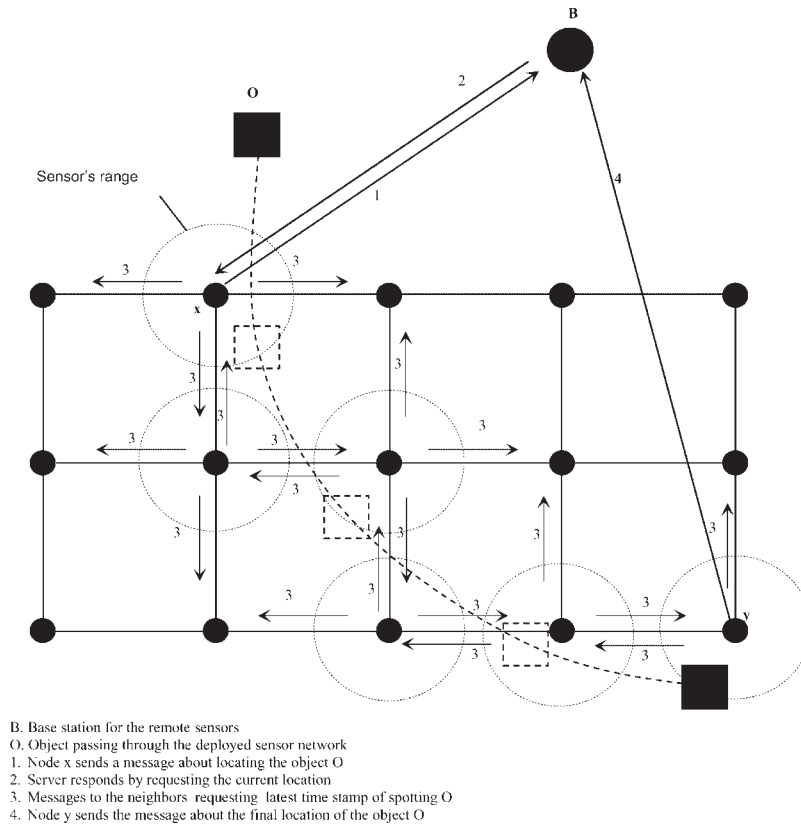


Fig. 1. Location tracking system.

sighting time as the answer to the query. Once returned, the base station updates the HN with the node that responded with the latest time stamp of the object. Next time, the same query is directed to its new HN.

4.1. Tracking Objects Using Maximum Radius/Velocity

Here, we propose another method of tracking objects over a sensor network that will track an object using parameters such as velocity. Since the velocity is very important parameter to define the object location in the given time frame, we calculate the radius, a threshold, which later defines the most probable area of object location. The nodes can calculate the optimal radius by calculating the difference between the last current velocity known and the maximum velocity possible. In case the last current velocity is not available, the minimum speed possible on that path can be used. This method works well in an environment such as child walking in a market place or on a highway system, where minimum and maximum speed is almost fixed.

The nodes that fall in the area (See Figure 2) will multicast the query to the entire sensor nodes. In the best case, this will save some short-term messages since based on the optimum radius based on the last velocity of the object, it can cover maximum number of nodes in that region. Thus, a node which has seen last the object, can respond to the base station. In the worst case, it may span the complete network.

Algorithm MaxRadius: Track target object over the sensor network.

Input: Lid: Logical identifier of the target object, Nid: Identifier of the sensor node.

V1: Velocity of object at time T1, V2: Velocity of object at time T2.

A: Multicast Area, SR: Minimum Distance between the two sensors

Output: T: Time of sighting of the target object by X.

Method:

- (1) $A = 3.14 * \left(\frac{(V1 - V2)}{(T1 - T2)} \right) * \left(\frac{(T2 - T3)}{POWER2} + SR \right) * POWER2$;
- (2) Query all nodes in A,

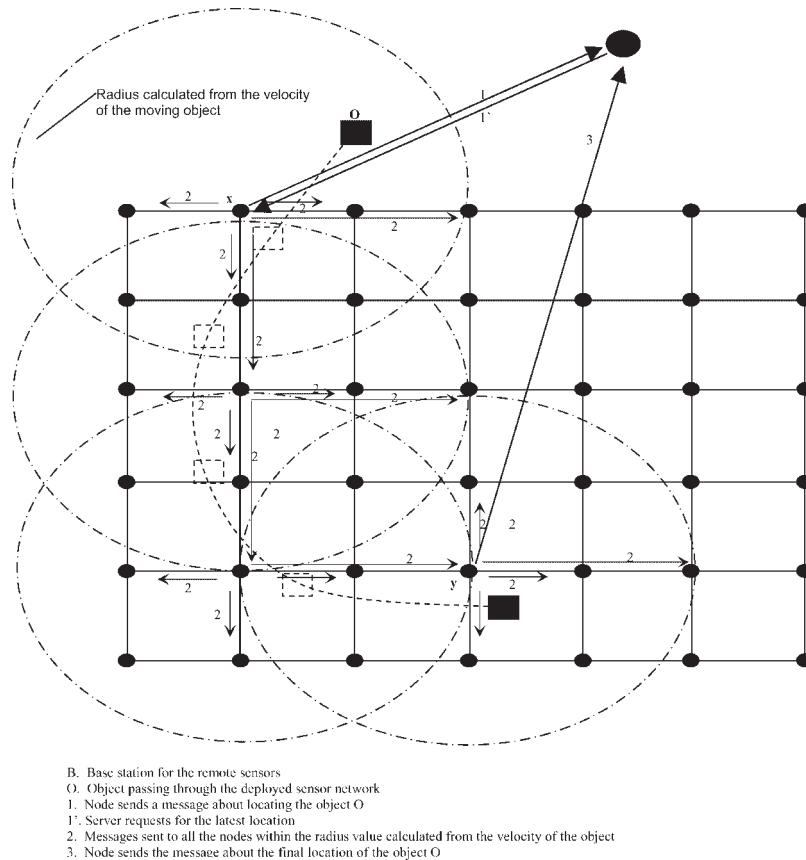


Fig. 2. Location tracking using maximum radius.

- (3) If no nodes in A has seen Lid sooner then send (Nid, T (Nid, Lid)) to the base station,
- (4) Else each node sends its T(Nid, Lid) if it is greater than T(HN, Lid) Else drop the Message,
- (5) Query the node with the latest T(Nid,Lid) to do TrackObject.

4.2. Memory Management

Here, we propose an algorithm, which effectively manages cache when the base station or any neighboring node sends a query to track objects. Whenever a sensor node references a memory location, the cache uses the unique object ID to select the set that should contain the cache data. Using our algorithm, the caching algorithm determines if the data is already present in one of the caches. If there is no data in its cache, then the sensor node lets the neighbor know by sending 'NULL.' If the sensor detects the new object and if cache is currently unused, pick the unused cache storage and enter the new data. If all cache lines are currently in use, then the cache controller must pick one of the cache lines and replace its data

with the new data. Ideally, we would like to keep the cache line which has not been referenced recently. Unfortunately, the cache is omniscient; they cannot predict the best one for replacement. However, by the principle of temporal locality, the object that has been referenced recently is likely to be referenced again in the very near future. A corollary to this is 'if a memory location has not been accessed in a while, it is likely to be a long time before the sensor node accesses it again.' Therefore, a good replacement policy that many caching controllers use is the 'least recently used' or LRU algorithm. The idea is to pick the cache line that was not most frequently accessed and replaces that cache line with the new data. An LRU policy is easily implemented in our cache system.

Algorithm Care-Less: Update the tracked object by a sensor node

Input: Lid: Logical identifier of the object

C: Cache Memory

Cptr: Cache Pointer

TS: Time Stamp

LS: Last Access

Output: C Update the cache line

Method:

- (1) While (C is *Null*)
 - Sensor node searches the location of the Search(Lid)
- (2) If (search(Lid) == *Null*)
 - *(Cptr) = Id;
 - Enter the TS of LS = TS against Cptr;
 - Cptr⁺⁺;

Else

- Calculate the least used cache line on LS using LRU;
- Enter the Id against the cache line;
- Enter the TS of LS = TS;

4.3. Randomized Algorithm for Short Range Communication

Given the highly probable interference, sampling over a time window per transmitter in the sensor node would be a good idea to estimate the correct location and also resolves the ambiguity in time stamp. For example, when the object is equidistance from the neighboring sensor nodes, there is always a high probability that neighboring nodes duplicate the same timestamp and thus, leads to inefficient memory management. Also, issues like carrier-sense-style channel to avoid collisions need intensive recourses and heavy computing, which the sensor nodes cannot sustain.

Instead, we handled these problems using a randomized algorithm. Rather than transmitting the signals continuously over time, each node will transmit the signals over an interval [R1, R2]. Thus, the broadcast of different sensor node are independent, which avoids the issues like time stamp ambiguity interference. In this system, the allocation of the timeslots is on a synchronous basis. The timeslots are shared on an equal basis in strict rotation (as shown in Figure 3).

Algorithm:

1. For each sensor node *m*, detect the time stamp of the object over the random time period [R1, R2] in ms.
2. The unique time stamp will be registered against the object in the cache.

5. Complexity and Performance Analysis

In this section, we will evaluate the performance of our algorithms and compare against the algorithm in

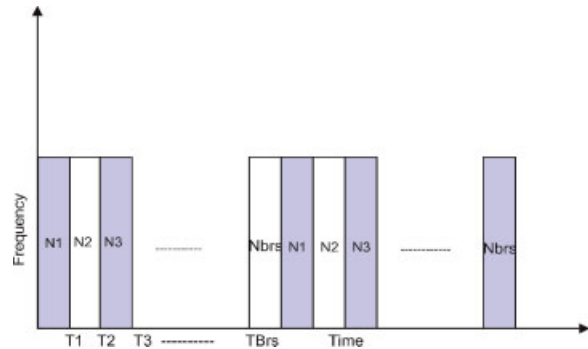


Fig. 3. Time slot allocation.

Reference [2]. Initially, we will analyze algorithms for the number of elementary target movements. Since any node in the sensor network can communicate to base station and vice versa, and the node that reported the latest time stamp for the query will be assigned as home node of the object and further query for the object will be directed to home node of the object. Initially, we like to discuss data that we generated randomly with a predefined structure. Basically, the data for simulation consists of two structured files:

1. Intrusion file: This file consists of intruder ID, the corresponding time stamp, and the sensor node IP address that object moved across the network.
2. The sensor file: This file consists of Node ID, time stamp, the message transmission path, and Intruder ID at the corresponding sensor node.

5.1. Analysis of Location Tracking (LT) Algorithm

In this algorithm, the communication between two neighbors is a short-range message. Moreover, the local computation within the sensor network with short messages will increase security and saves bandwidth against communication on the common channel to the base station.

On receiving a query from the base station to HN, the search will lead to an *n*-array tree, where the order of the tree depends on the neighbors of HN. Since in the best case, assuming that object would not move in subsequent query, the number of messages would be limited to the neighbors. Hence, in best case we could track the object using *Nbrs* Where *Nbrs* = number of neighbor nodes of the queried node (In our case we assume that the number of neighbor nodes is *Nbrs* for the each node in the grid).

As the object moves during a query, the tracking would not be limited to neighbors of the HN. The HN

starts the process of following the trail for the queried target object. Once the target object moved out of range of a sensor node, it must be in the range of one of its nbr node in the grid. By finding the particular neighbor who spotted the target object last, will determine the next step in the trail of the target object. The trail stops when either the target object is still in the range of this existing node, or the elected node is an edge node and the target object has left the grid of sensors. Such a node then responds back to the base station giving its identifier and sighting time as the answer to the query. The base station will update the HN corresponding to the object and the next query for the object would be directed to new HN. Thus, a careful examination tells us that the search for the object follows an n -order tree. Assuming a uniform grid with Nbrs number of neighbors, the Figure 4 below depicts an n -array search tree.

In the worst scenario, the search cost is the number of messages exchanged which is given by:

$$O(\lceil \log_{Nbrs}(N \times N) \rceil) \quad \text{where}$$

Nbrs is the number of neighboring nodes of the queried node and N is the number of nodes in row/column in a uniform square grid. Hence, the number of total messages would be:

$$O(\lceil \log_{Nbrs}(N \times N) \rceil \times Nbrs)$$

Here, we assume that the number of neighbor nodes is Nbrs for the each node in the grid

5.2. Analysis of Maximum Radius (MR) Algorithm

In Maximum radius (MR) algorithm, the nodes will multicast the query to the sensor nodes depending on the object's velocity that fall in the area. In the best case, this will save some short-term messages as based on the last tracked velocity and the maximum possible velocity of the object, it can cover the maximum nodes at the edge of the circle. Thus, a node which has seen the object last can respond to the base station. In the worst case, it may span the complete network.

In Figure 5, the base station contacts the HN X, and later X defines the radius depending on the behavioral parameter such as velocity of the object. Then, the node X contacts the nodes within the radius based on the maximum velocity constraint in that area and the last tracked velocity to get the maximum radius, and so on. Here in our example, the node Y will be the HN after trail ends. The best case would be finding the object in the very second multicast, that is $Nd_{M1} + Nd_{M2}$ where Nd_{M1} = Number of the nodes in first multicast and Nd_{M2} = Number of the nodes in second multicast.

In the worst scenario, the HN will span the entire network. Hence, the number of messages would be: $Nd_{M1} + Nd_{M2} + Nd_{M3}, \dots, + Nd_{Mn}$

In this case, we assume that the home node will not send the messages that has received for a given target object.

Here, we have simulated algorithms in a uniform grid with unit distance between two neighbor nodes and Nbrs as the neighboring nodes for each node in the grid. As far as the LT and Algorithm [2] are concerned they would not deviate much from

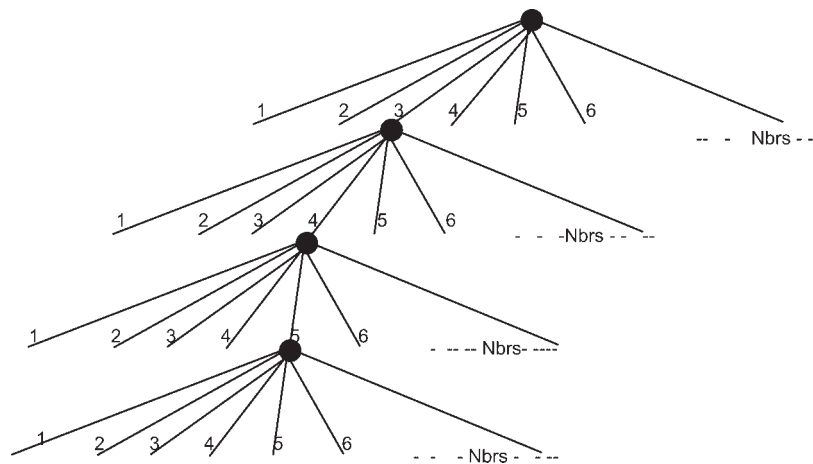


Fig. 4. N -array search tree.

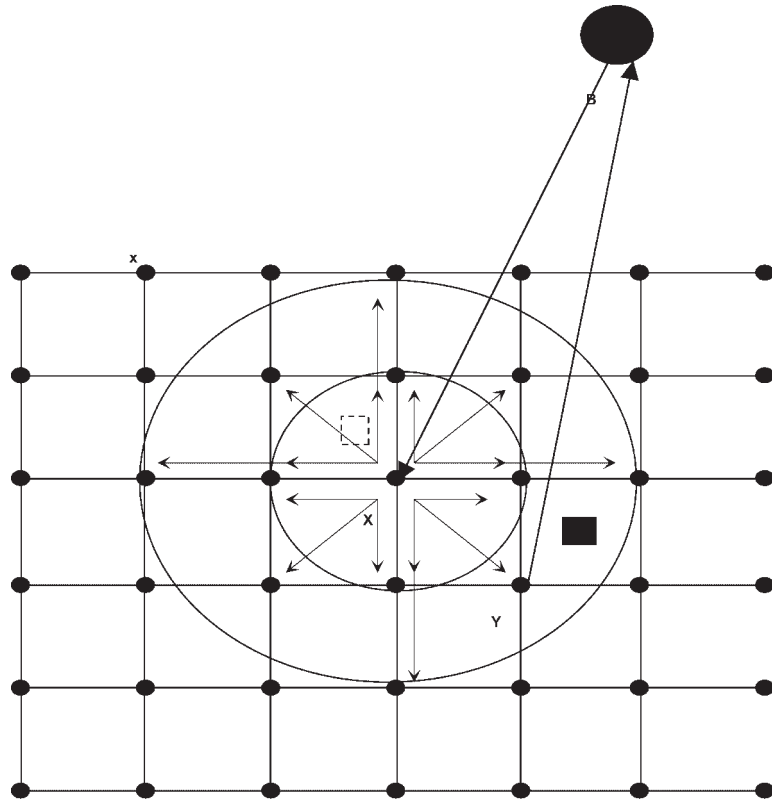


Fig. 5. Maximum radius algorithm analysis.

generalization until Nbrs varies for each node in the grid. The Figure 7 depicts the worst and best case for both LT and Algorithm [2].

In MR algorithm, the analysis varies on different parameter such as velocity, grid structure, and other object behavioral patterns. For the performance analysis of MR and its comparison with the other algorithm [2], we assumed a uniform grid structure as above and predict the number of nodes in each circle. Let us assume that the distance between the neighboring nodes as one unit. The number of nodes within radius K (an integer) is as follows:

Number of nodes at distance $K = 1$

Number of nodes at distance

$$(K - 1) \cong \left\lfloor 2 \times \sqrt{K^2 - (K - 1)^2} \right\rfloor$$

Number of nodes at distance

$$(K - 2) \cong \left\lfloor 2 \times \sqrt{K^2 - (K - 2)^2} \right\rfloor$$

.....

Number of nodes at center = $\lceil 2 \times K \rceil$

NoM	LT	Algorithm[2]	MR algorithm
Best	$O(Nbrs)$	$O(N+Nbrs)$	$O(Nd_{M1} + Nd_{M2})$
Worst	$O(\lceil \log_{Nbrs}(N \times N) \rceil \times Nbrs^2)$	$O(\lceil \log_{Nbrs}(N \times N) \rceil \times Nbrs^2)$	$O(Nd_{M1} + \dots + Nd_{Mn})$

Fig. 6. Complexity of different algorithms.

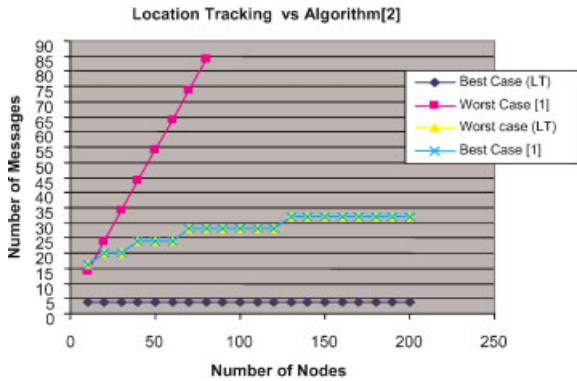


Fig. 7. Location tracking v/s algorithm [2].

Hence, the number of nodes in a circle with radius K

$$\cong \left\lceil 2 \times \left(2 \times \sum_{X=K-1}^1 \sqrt{K^2 - X^2} \right) \right\rceil + 2 + \lceil 2 \times K \rceil$$

The graph in Figure 8 depicts the number of messages versus the radius of multicast circle. Here in the graph we see that the number of messages increased as the radius of the circle increases (we have normalized the number of messages using log value shown as ‘ln’ in the figure). Thus, expected energy consumption increases with the increase in the radius. Although, this algorithm will have better results with very highly dense grid compared to the other two algorithms, but there are other behavioral aspects about the object such as object has stopped or accelerated the speed suddenly, or the grid structure has

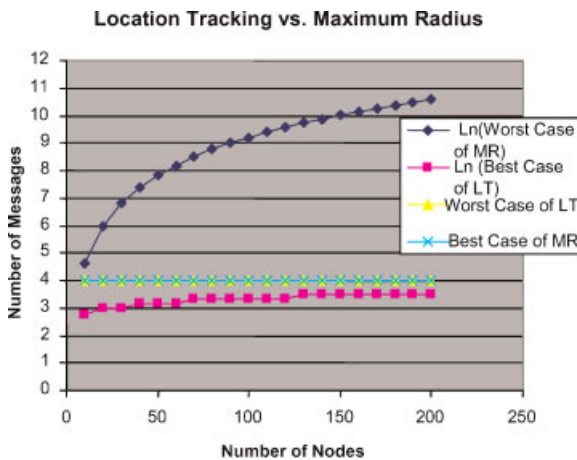


Fig. 8. Performance analysis of MR algorithm.

changed, which can impact the performance. Though in the worst case, MR has more number of messages, but its response time will be much faster than others.

6. Conclusions and Future Work

The two algorithms given in this paper aim at tracking a moving object in a sensor network. The sensor that is currently holds the latest time stamp of the object acts as the home node and communicates with the base station in the first algorithm. In the second, the minimum and maximum velocity is used to predict the radius of flooding, thus avoids the complete flooding of the network. The query shipping approach allows for smaller messages to be transmitted, thus saving energy and channel usage against the conventional data shipping. Another major advantage of our model is that, the long-range messages between base station and the nodes are minimized by the local computation within the sensor nodes of the network; this further reduces the power usage in the nodes. Collectively, the query shipping approach and minimizing long range messaging help in efficiently utilizing the sensor’s limited power resources. The location of the object can be retrieved by only two long range messages, base station query, and the corresponding time stamp message in addition to short range messages to all the neighbors for querying the latest time stamp. Finally, the node with the latest time stamp of the object reports to the base station will become the home node for the object. The simulation study shows the advantages of the two algorithms in terms of complexity and the experimental evaluation. In future, we are implementing the algorithms using sensor nodes and will compare the results with the experimental data.

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