Modeling the Connections of Dynamic Sensor Fields Based on BT-Graph

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Abstract

In this article, we propose a new approach to model and optimize the dynamic sensor field for both internal network connections and LEO satellite connection based on BT-Graph. Due to the shift of LEO satellite's orbit at each revolution, a dynamic sensor field (DSF), which is able to redetermine its gateways, is suitable to improve successful data communications. It is convenient to present a DSF as a BT-Graph that aims to utilize optimization algorithms. Parallel search algorithms are also deployed for more efficient execution time when DSFs consist of a significant large number of nodes. The simulation experiments are performed on an abstract forest fire surveillance network to validate our proposed approach.

1. Introduction

Wireless Sensor Network (WSN) [1] is known as a network of sensors cooperatively operating in order to surveillance or collect the environmental parameters. Although the short transmission range can be compensated by applying a mesh topology, it would be economically infeasible to deploy in large geographic areas or behind obstacles (mountains, oceans, ...). To overcome these disadvantages, satellite-based wireless sensor networks have emerged as a promising technology with various applications.

Because the orbit of a LEO satellite shifts in westward direction around the polar axis at each revolution [12], the meeting points of a gateway on the Earth's surface and the LEO satellite will be changed over time. With a static sensor field, it can be occasionally unsuccessful in communication with the LEO satellite because the meeting time is not enough for data exchange [12]. Therefore a dynamic sensor field (DSF) [1], which has the ability to redetermine its gateway to adapt with the shifts of LEO satellite's paths, is proposed to improve the connection time.

It is necessary to choose proper gateways for the longest length of connection time. Consequently, the connections between a LEO satellite and a dynamic sensor field should be presented by a graph-based model because it is convenient to determine number of neighbors in the satellite's communication range and then apply the optimization algorithms. Moreover, BT-Grap model [2] based on balltree structure is efficiently support not only for searching the range nearest neighbors (RNN) but also finding the shortest path to a given target node [4] [18] [20]. In this article, we propose a new approach, namely dynamic sensor field optimization model based on BT-Graph (BT-DYNSEN), to model and optimize the DSF in communication with LEO satellite.

In addition, with the development of Compute Unified Device Architecture (CUDA), general purpose Graphics Processing Unit (GPGPU) is widely used to implement parallel algorithms, which can speed up the large and complex processing tasks [7] [8]. Nonetheless, serial graph algorithms are hard to be parallelized on GPUs due to irregular memory access and huge

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memory-intensive operations. For this reason, we also propose parallel algorithms to improve search speed based on BT-Graph model using NVIDIA CUDA GPUs. In our work, the experimental results were deployed on parallelization of RNN algorithm and Dijkstra's shortest path search algorithm.

The rest of this article is organized as follows. In section 2, we overview related works. Section 3 presents the graph-based model for a DSF based on BT-Graph (BT-DYNSEN). How to optimize the DSF for satellite connection is presented in section 4. Implementation of BT-DYNSEN is described in section 5. Section 6 gives the simulation experiments on an abstract network for forest fire surveillance in Vietnam before a conclusion is drawn.

2. Related work

The LEO satellite communication can be used in mobile satellite communications, surveillance the Earth surface, geological surveys, so on [1] [16] [19]. In the last decade, researches on communication services provided by LEO satellites have focused on several main directions such as optimizing the mechanics, interconnections, electric circuits, power supply. On another approach, many surveys show the research topics in design trajectory, handover traffic and constellation, as well as design protocols, radio frequencies, onboard transceivers and antenna designs [13] [14] [1] [19]. However, the direct radio links between sensor fields and LEO satellites are not considered in literature. In recent years, it emerges as an attractive topic because of the current innovation solutions such as LoRa Semtech and solutions from vendors QB50 [11].

Furthermore, graph-based model has emerged as well approach to present the structure and elaborate the performance of wireless sensor networks [3]. For example, random geometric graph [20] was used to determine the probability of the whole network being connected. Secure communications between large number of sensor nodes in WSNs can be elaborated on expander graph [25] and finding transmission path in network was performed based on Pascal graph [5]. Furthermore, hyper-graph [24] was utilized to support for reducing the transmission energy consumption and improving the fault-tolerant ability of the system.

Additionally, in recent years parallelization on GPU platforms is an emerging strategy to improve significantly performance speed [9]. When the graph consisting of a vast number of nodes, it is necessary to process in parallel based on NVIDIA CUDA technology to improve search speed [10]. In the next section, we introduce BT-DYNSEN model for optimizing the DSF for the connection with LEO satellite.
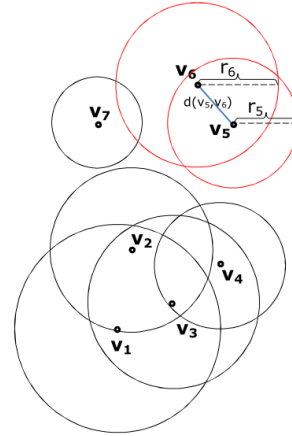

Figure 1. A dynamic sensor field in which the communication ranges of sensor nodes are indicated by the radii of balls.
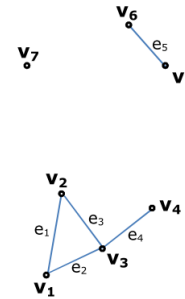

Figure 2. A BT-Graph of the dynamic sensor field with 07 vertices $V = \{v_1, v_2, v_3, v_4, v_5, v_6, v_7\}$ and 5 edges $E = \{e_1 = e(v_1, v_2), e_2 = e(v_1, v_3), e_3 = e(v_2, v_3), e_4 = e(v_3, v_4), e_5 = e(v_5, v_6)\}$.

3. BT-DYNSEN

3.1. BT-DYNSEN model of DSF

BT-DYNSEN model of a dynamic sensor field (DSF) [1] based on BT-Graph [2] is a graph G(V,E). In this graph, set of vertices $V = \{v_i\}$, $i = 1..n$ corresponds to sensor nodes and set of edges, $E = \{e_j\}$, $j = 1..m$ are connections between the nodes with associated weight functions $W = \{w_j\}$, $j = 1..m$. The value of each w_j is given by Euclidean distance $d(v_i, v_j)$, $i \neq j$. Additionally, $R = \{r_i\}$, $i = 1..n$ are the radii of the communication ranges of nodes [2] [4] [6]. An edge is established if and only if the distance between two nodes is less or equal to the minimum value of their communication radii, $d(v_i, v_j) \leq \min(r_i, r_j)$, $i \neq j$. Note that the terms *node* and *vertex* are used interchangeably in this article as a matter of convenience.

In Figure 1, for an example, a pair of vertices (v_5, v_6) has communication ranges r_5 and r_6 respectively. Because the distance between v_5 and v_6 is less than r, $d(v_5, v_6) < r$, so there exists an edge e_5 connecting them as can be seen in Figure 2. Similarly, the others edges

Table 1. An example of satellite connection data with three rows: *Connection number*, *Connection vector* and *Time vector*.

Connection number	1	2	3	...
Connection vector	v_1	v_3	v_4	...
Time vector	t_1	t_2	t_3	...

of this graph namely $e_1 = e(v_1, v_2), e_2 = e(v_1, v_3), e_3 = e(v_2, v_3), e_4 = e(v_3, v_4)$ could be established. There are $2^n - 5$ edges between a pair of nodes of this graph that are not existed due to inadequacy of the condition. The vertex v_7 is isolated because all distance values between it and the other nodes are inadequate to the condition.

3.2. BT-DYNSEN model of LEO satellite connection

The connections between a LEO satellite and a dynamic sensor field are also described in a BT-Graph. Let $V = \{v_i\}$, i=1..n, is a set of sensor node coordinates in a DSF. In this scenario, connection time is defined when any sensor set [1] of the DSF under the satellite coverage. The LEO satellite communication range is considered as a circle whose center is sub-point on the ground (sub-satellite point), s. It is noted that sub-satellite point, s, is where on the ground the straight line connecting the center of the Earth and the satellite meets the Earth's surface [12]. The associated BT-Graph consists of $n + 1$ vertices $P = \{V, s\} = \{v_1, v_2, ..., v_n, s\}$. If a node is within the communication range of the satellite, there exists an edge with weight given by the Euclidean distance between them. Otherwise edge weight is set to infinity. Consequently, the number edges of the graph are $m + n$ by adding n new edges $C = \{c_1, c_2, ..., c_n\}$ with $c_i = c(s, v_i)$, $i = 1..n$. The weights of the n new edges are denoted by $Z = \{z_1, z_2, ..., z_n\}$. In this case the set of edges is $R = \{E, C\} = \{e_1, e_2, ..., e_m, c_1, c_2, ..., c_n\}$ and the set of corresponding weights is $Q = \{W, Z\} = \{\{w_k\}, \{z_i\}\}$ with $k = 1..m$, $i = 1..n$. Hence the BT-DYNSEN model for LEO satellite connections is presented by a graph G(P,R) with weight functions Q.

Furthermore, during the connection time only one sensor node (vertex) of a dynamic sensor field is chosen to connect with the sub-satellite point (center vertex) at one time [1]. A number of different nodes could be chosen based on the value of edge weights at different times. To manage the connections, the name of chosen node is kept in *Connection vector* [1] and the corresponding time is saved in *Time vector* [1] (as in Table 1).

Table 1 shows that in *connection 1*, center vertex s connects with v_1 at time t_1. In a similar way, in *connection 2* at time t_2 and *connection 3* at time t_3, v_4 and v_2 are chosen to connect with s respectively.

For instance, Figure 3 shows three graphs of the dynamic sensor field in three different connections with the center vertex s (a sub-satellite point) at different
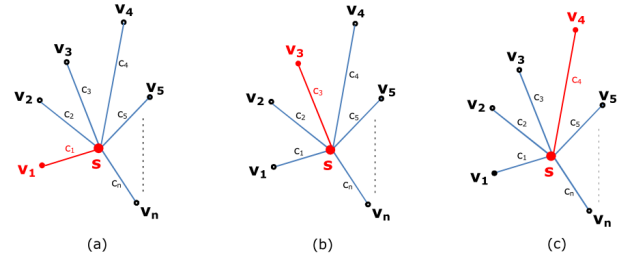

Figure 3. The BT-Graph of a dynamic sensor field in three different connections (solid red lines) at times t_1, t_2 and t_3.
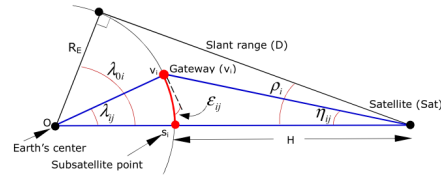

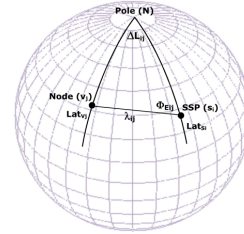**Figure 4.** Gateway of sensor field geometry [12].

Figure 5. Relationship between sensor node of a DSF and sub-satellite point [12].

times t_1, t_2 and t_3. At time t_1 (Figure 3(a)), vertex v_1 is chosen and edge c_1 is established. Similarly, in Figure 3(b) and Figure 3(c) vertex v_3, v_4 are chosen for connections that leads to corresponding edges c_3, c_4 are established at time t_2 and t_3 respectively.

3.3. Compute the connection time

In this section, we describe the way to calculate connection time, t_{ij}, between a LEO satellite and a gateway of DSF, v_j [1] [12] [15]. Similarly, the calculation could be applied for all other nodes. Note that every node of the sensor field is assumed as a gateway for the connection with the satellite in calculating the values of connection time.

First, it is necessary to define the angles and related distances between satellite, a gateway on the ground and the Earth's center. The parameters are indicated on Figure 4. For angular radius of the spherical Earth, ρ_i, and the angular radius λ_{0_i} can be found from relations

$$\sin(\rho_i) = \cos(\lambda_{0_i}) = \frac{R_E}{R_E + H} \qquad (1)$$

$$\rho_i + \lambda_{0_i} = 90 \ deg \qquad (2)$$

3

where $R_E = 6378.14$ km is the Earth's radius and H is the altitude of the satellite above the Earth's surface.

With the coordinates of a sub-satellite point, s_i ($Long_{s_i}, Lat_{s_i}$) along each satellite's ground track and a sensor node, nodes v_j of the DSF ($Long_{v_j}, Lat_{v_j}$), and defining $\Delta L_{ij} = |Long_{s_i} - Long_{v_j}|$, the azimuth, $\Phi_{E_{ij}}$, measured eastward from north, and angular distance, λ_{ij}, from the sub-satellite point to the sensor node (see Figure 5) are given by

$$cos(\lambda_{ij}) = sin(Lat_{s_i})sin(Lat_{v_j}) + \\ cos(Lat_{s_i})cos(Lat_{v_j})cos(\Delta L_{ij}) \tag{3}$$

$$cos(\Phi_{E_{ij}}) = ((sin(Lat_{v_i}) - \\ cos(\lambda_{ij})sin(Lat_{s_i}))/(sin(\lambda_{ij})cos(Lat_{s_i})) \tag{4}$$

where $\Phi_{E_{ij}} < 180\ deg$ if v_i is east of s_i and $\Phi_{E_{ij}} > 180\ deg$ if v_i is west of s_i. Consider triangle Os_iv_j (in Figure 4), the distance, d_{ij}, between s_i and v_j can be found using the law of cosines:

$$d_{ij}^2 = R_E^2(1 - cos(\lambda_{ij})) \tag{5}$$

The *connection time*, t_{ij}, is given by

$$t_{ij} = \left(\frac{P}{180\ deg}\right)arccos\left(\frac{cos(\lambda_{ij_{max}})}{cos(\lambda_{ij_{min}})}\right) \tag{6}$$

where P is the *orbit period* in minutes. From Equations (5) and (6), the communication duration, t_{ij}, strongly depends on how close the nodes v_j is to the sub-satellite points s_i, the distances d_{ij}, along the ground track on any given orbit pass [12].

3.4. Constraint

The altitude of a LEO satellite must be in range from 275 km to 1400 km due to atmosphere drag and Van Allen radiation effects [12]. Besides, the experimental results were announced by High Altitude Society in United Kingdom that LoRa SemTech transceivers can communicate in distance up to 600 km in environment without any obstacle and 20-40 km in urban area [22]. Based on these factors, maximum communication range for all nodes in long range sensor fields is 40 km in this work. LEO's satellites are chosen in our experiments must have the orbit altitude less than 600 km. Furthermore, the satellite's relative speed over a fixed point on the Earth's surface must be around 7.5 - 8.0 km/sec [12]. The speed of the satellite is calculated by

$$v = \sqrt{\frac{Gm_E}{R_E + H}} \tag{7}$$

Equation 7 shows that the speed of the satellite in orbit is in inverse proportion of its altitude [12]. Where

G is universal gravitational constant ($G = 6.67\text{x}10^{-11}$ Nm^2/kg^2) and m_E is the mass of the Earth ($m_E = 5.98\text{x}10^{24}$ kg). With orbit altitude of satellite in range 300-600 km, the speed of satellite on orbit must be in range 7.56-7.73 km/sec.

4. Optimization method

4.1. Verify the connectivity of DSF

In this work, top down construction algorithm is chosen in order to build balltree structure for verifying the network connectivity of the DSF. In this manner, the time complexity can be found in $O(nlog^2 n)$ [6].

Algorithm 1 Verify the network connectivity of a DSF

1: **procedure** VERIFYCONNECTIVITY(D)
2:     **if** D.hasAPoint() **then**
3:         Create a leaf B is a point in D
4:         Return B
5:     **else**
6:         Let c is the dimension of greatest spread
7:         Let L, R are the set of points lying to 2 subsets
8:         Let r is a radius of ball
9:         Create B with two children:
10:             B.center ← c
11:             B.radius ← r
12:             B.leftChild ← balltreeConstruct(L)
13:             B.rightChild ← balltreeConstruct(R)
14:     **end if**
15: **end procedure**

Top down algorithm to construct the balltree is a recursive process from top to down. The process at each step is to choose the split dimension it and then split the set of values into two subsets. First, it is necessary to choose the point in the ball which is farthest from its center, p_1, and then choose a second point p_2 that is farthest from p_1. It is followed by assigning all points in the ball to closest one of two clusters corresponding to p_1 and p_2. Finally, the centroid of each cluster is calculated and the minimum cluster radius for enclosing all the points are determined. This process is stopped only when the leaf balls contain just two points. Note that D denotes current considered balltree structure and B denotes leaf node defined after each process.

4.2. Determine the sensorsets corresponding to each sub-satellite point

Defining a sensorset [1] based on BT-DYNSEN is to find k nearest sensor nodes (neighbors) under the satellite's coverage area at each sub-satellite. It is a depth-first traversal algorithm for traversing tree or graph data structures, starting with the root node. The value of Q

is updated during search process. At each considered node B, Q obtains k points which are nearest query q as following algorithm.

Algorithm 2 Determine a sensorset

1: **procedure** DETERMINESENSORSET
2:     **if** balltreeNode.isALeaf **then**
3:         **if** d < query.range **then**
4:             Write name of balltreeNode to file
5:         **end if**
6:     **else**
7:         **if** d < (query.range + radiusNode) **then**
8:             left ← balltreeNode.leftChild
9:             right ← balltreeNode.rightChild
10:            rnn: left, query
11:            rnn: right, query
12:         **end if**
13:     **end if**
14: **end procedure**

There are two different cases in *BT-DYNSEN* algorithm as follows. If current considered node is a leaf node and the distance from query point q to B is less than r ($d < r$), the obtained result is updated by adding B into Q. Otherwise, if B is not a leaf node and the distance from query point q to B is less than the total of r and the radius of B ($d < r + B.radius$), it is necessary to perform recursive algorithm for the two child nodes of a parent node B: left-child and right-child.

4.3. Select the gateways of DSF

If a DSF, V, has n nodes, there are $2^n - 1$ connection items, $G = \{g_l\}$, l=1..($2^n - 1$). It is necessary to find out a set of proper nodes which play as gateways of DSF to provide the longest length of time for the connection. To do this, the *association analysis algorithm* [18] is applied. In this way, a DSF is represented in a binary format, where each row corresponds to a connection item and each column corresponds to a node. A node value is one if the node appears in a connection item and zero otherwise. Weight of a connection item determines how often a connection item is applicable to a set of connection items. The weight of a connection item, $g_i \in G$, can be defined as $w(g_i) = |\{g_l \mid g_i \subseteq g_l, g_l \in G\}|$, where the symbol $|.|$ denotes the number of elements in a set.

The algorithm for selecting the gateways of the DSF is briefly presented in following list.

For example, a DSF with 4 nodes, $V = \{v_1, v_2, v_3, v_4\}$ is shown in Figure 6. There are 4 sensorsets $A_1 = \{v_1\}$, $A_2 = \{v_1, v_2, v_3\}$, $A_3 = \{v_2, v_3, v_4\}$ and $A_4 = \{v_3, v_4\}$. The weights of connection items are presented in Table 2. The connection with the highest weight corresponds to the least number of times required to change the gateways. It leads to the connection duration time of the

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Algorithm 3 Select the gateways of a DSF

Input: list of sensorsets
Output: gateways of DSF
1: C_k: candidate sensors (size k)
2: L_k: set of selected gateways (size k)
3: **procedure** SELECTGATEWAY
4:     $L_1 \leftarrow \{weight1 - sensorsets\}$
5:     **for** (k=2;Lk.count>0;k++) **do**
6:         $C_{k+1} \leftarrow$ generate candidate sensors L_k
7:         **for each** *transaction t ∈ data* **do**
8:             Increment count of the candidate sensors in C_{k+1} that contained in t
9:         **end for**
10:        $L_{k+1} \leftarrow$ candidate sensors in C_{k+1}
11:     **end for**
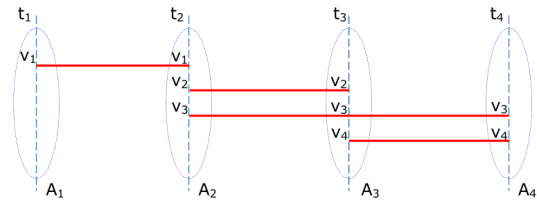12:   Write counted frequent of all sensors L_k to file
13: **end procedure**

Figure 6. The connections between a 4-node DSF with a LEO satellite.

Table 2. The weights of connection items in a DSF with 4 nodes.

Connection items	Weights
(v_1, v_2, v_4)	1
(v_1, v_2, v_3)	1
(v_1, v_3)	3
(v_1, v_3, v_4)	1

connection is longest. Because the weight of connection item (v_1, v_3) is highest, this connection is chosen.

4.4. Find shortest path for data dissemination

The problem of finding shortest path from each sensor node of DSF to the gateway can be solved by a graph search method. The algorithm is presented as Algorithm 4.

The algorithm proceeds in three following steps. First, the balltree graph-based model, G, is constructed. At the next step, the weight matrix M for edges connect each pair of vertices (sensor nodes) in V is computed based on their coordinates. For v_i, v_j are two different vertices in V, in case of $v_i \equiv v_j$ the edge weight is 0. If $v_i \not\equiv v_j$, the edge weight is given by $d(v_j, v_j)$ if $d(v_j, v_j) \leq r$, otherwise it is infinity, ∞. Finally, the shortest path from q to e in the weight matrix is figured out. Note

Algorithm 4 Find shortest path from each sensor node of DSF to the gateway

```
 1: procedure FINDSHORTESTPATH
 2:     Init matrix M
 3:     S ← Start point
 4:     T ← End point
 5:     d is an array
 6:     free is an array
 7:     trace is as array
 8:     d[S] ← 0
 9:     while TRUE do
10:         u ← -1
11:         min ← INFINITE
12:         for v ← 0 to n-1 do
13:             if free[v] AND (d[v] > (d[u] + M[u, v]))
    then
14:                 d[v] ← d[u] = arr[u][v];
15:                 trace[v] ← u;
16:             end if
17:         end for
18:         if (u=-1) OR (u=T)) then
19:             Break
20:         end if
21:         free[u] ← false
22:         for (v ← 0 to n-1 do
23:             if free[v] AND (d[v]>(d[u]+M[u,v])) then
24:                 d[v] ← d[u] = arr[u][v];
25:                 trace[v] ← u;
26:             end if
27:         end for
28:     end while
29: end procedure
```

that q is starting point (a sensor node) and e is the destination point (the gateway of DSF).

4.5. Parallelization of RNN algorithm to determine sensorsets

As mentioned in section 4.2, range nearest neighbors (RNN) search method on BT-Graph can be used to determine the set of nodes in a sensorset associated with each position of satellite along ground track. In this section, a greedy algorithm is proposed to nearest neighbors search running on CUDA (namely RNN-CUDA as shown in Algorithm 5). The process of this greedy algorithm consists of two part. One part is parallel processes on GPU to compute distances from query the point to all nodes of a dynamic sensor field. Another part runs on CPU to sort the obtained distances in ascending order and then pick up k neighbor nodes associated with shortest distance values.

Algorithm 5 RNN-CUDA

```
 1: ▷ Host program executed on CPU
 2: Load data into global memory
 3: Assign the value of k
 4: Copy data from CPU → GPU
 5: ▷ Parallel programs executed on GPUs
 6: Compute distances from query point, q, to all nodes
    v ∈ V
 7: Copy data back from GPU → CPU
 8: ▷ Host program executed on CPU
 9: Sort the distance values as a ascending list
10: Pick up k nearest neighbor nodes associated to k
    first value of the list
11: Release GPU memory
```

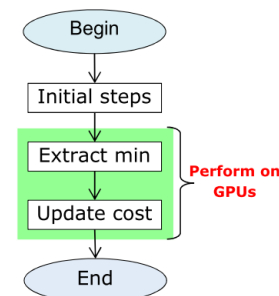

Figure 7. The flo chart of parallel Dijkstra algorithm for searching BT-Graph model.

4.6. Parallelization of search algorithm to fin the shortest path for data dissemination

To parellelize Dijkstra algorithm, we propose that each edge of BT-Graph is handled by a thread of CUDA. Figure 7 depicts the flow chart of parallel Dijkstra algorithm for searching BT-Graph model in which two procedures, *Extract min* and *Update cost*, are performed on NVIDIA GPUs. The parallelization of Dijkstra's algorithm is illustrated in Algorithm 6.

5. BT–DYNSEN implementation

We have developed the BT-DYNSEN tool in C/C++, that enables to model and optimize the connection time between a dynamic sensor field with a LEO satellite. GPredict [17] is used to provide the information about BEESAT-3 satellite's path. Besides, NetGen tool [21] is utilized to generate the abstract network of a 50-node dynamic sensor field from geographic data provided by Google maps service. The obtained result are the nodes of the DSF which should be configured as gateways for the best connection duration time and the shortest paths for data dissemination from each node to these gateways.

Algorithm 6 DijkstraCUDA

1: **Step 1:**
Initialize weight matrix (input vectors), start point and stop point

2: **Step 2:**
Allocate vectors in device memory corresponding to each edge of BT-Graph

3: **Step 3:**
Copy vectors from CPU (host) memory to GPU (device) memory

4: **Step 4:**
- Search for a free vertex u that has the least cost from the starting vertex S to u
  + If no vertex u is found, either it exists a path or not
  + Otherwise, from u we continue to consider the other free vertices
- Use CUDA to specify the thread for searching the paths between the vertex u and the others

5: **Step 5:**
Copy results from GPUs memory to CPU memory

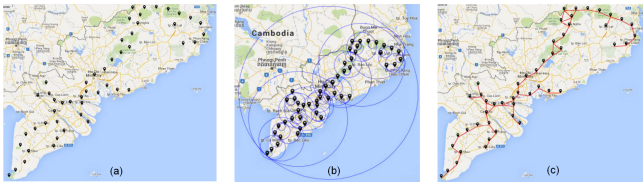6: **Step 6:**
Free device memory and host memory

Figure 8. Verify dynamic sensor network connectivity (a) A 50-node dynamic sensor network, (b) Balltree structure of the DSF, (c) Connectivity of the DSF.

6. Experiment

6.1. Experiment 1: Verify network connectivity

The connectivity of the 50-node dynamic sensor network was verified by applying BT-DYNSEN as shown in Figre 8(a). Figure 8(b) depicts the balltree structure of this DSF in which there are 49 balls with radii from 10.124m to 321.165m. BT-Graph model then was utilized to ensure the full connectivity of all network nodes, the radius 30.497m was then chosen as shown in Figure 8(c).

6.2. Experiment 2: Determine the sensorsets

BT-DYNSEN tool was employed in determining sensorsets corresponding to sub-satellite points during visiting time. The map in Figure 9 shows the sensorset was determined with sub-satellites of BEESAT-3 at (latitude: 14.00, longitude: 102.48) in orbit 12794. Sub-satellite and satellite coverage were indicated by solid
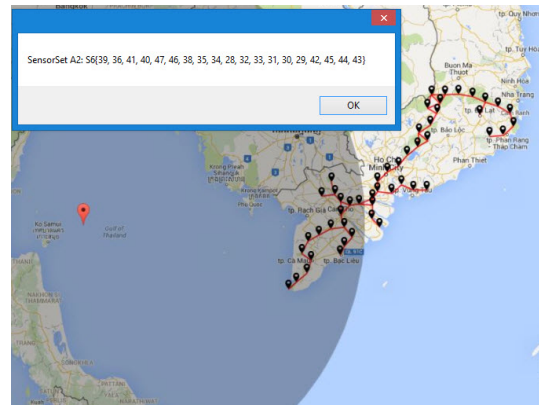

Figure 9. Determine the sensorsets corresponding to each sub-satellite along the groundtrack of BEESAT-3 in orbit 12794.
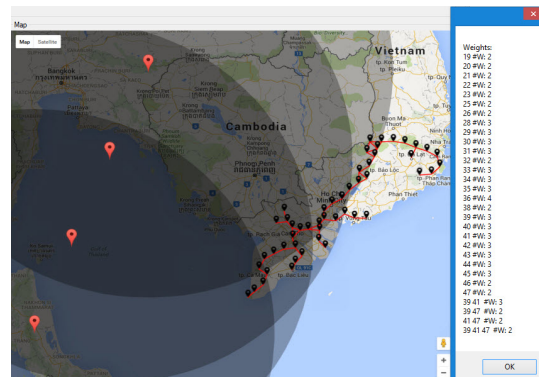

Figure 10. Select a set of nodes which play as gateways of the DSF according to the weights of connections.

red square and red circle repectively. The sensorset consists of 19 sensor nodes which were under the satellite's coverage area (the shadow area). There are four sensorsets were created along the BEESAT-3's ground track in orbit 12794.

6.3. Experiment 3: Select the gateways

With each sensorset, a subset of connections is established. The weights of each connection in subset are then computed. A set of connection items is created by combining these subsets. The best connection is chosen based on the weights of connection items. For instance, in Figure 10 node v_{36} was chosen as the gateway of the 50-node DSF to connection with BEESAT-3 satellite in orbit 12794 because its weight is highest in sensor nodes.

6.4. Experiment 4: Optimize data dissemination

To ensure sensing data to be collected from all sensor nodes and then sent before the satellite leaving, it is necessary to find the shortest path from each

Figure 11. The shortest path for data dissemination from node v_{50} to gateway node v_{36}.

Table 3. Comparison on the execution time of RNN BT-Graph and RNN-CUDA.

	N=10000, Q=10000	N=100000, Q=100000
RNN BT-Graph	179.46	17898.13
RNN-CUDA	123.47	5440.29

sensor nodes to the gateway. The interconnection weights within the DSF are geographic distances between each pair of nodes that were carried out by applying BT-DYNSEN model. Figure 11, as an example, illustrates the chosen path (the bold blue line) for data dissemination from sensor node v_{50} to the gateway node v_{36}.

6.5. Experiment 5: Parallelization RNN algorithm for determining sensor sets

In order to make the comparison between the parallel algorithms and serial ones for searching BT-Graph, we have run the simulations on a computer which is equipped with a INTEL® Core i3-3220 3.3GHz processor, 4GB DDR3 SDRAM and a NVIDIA GeForce GTX660 graphic card. In our experiment, a comparison about BT-Graph's search speed is obtained by performing both sequential algorithm on CPU and parallel algorithm on GPUs. On each algorithm with the same input data, which are generated randomly by program, experimental program is executed in 10 times for getting the average value of execution time.

The results is illustrated in Table 3, where N is the amount of initial data points and Q is a number of query points. It shows that the search speed of RNN-CUDA be improved around 2.37 folds compared with that of RNN BT-Graph.
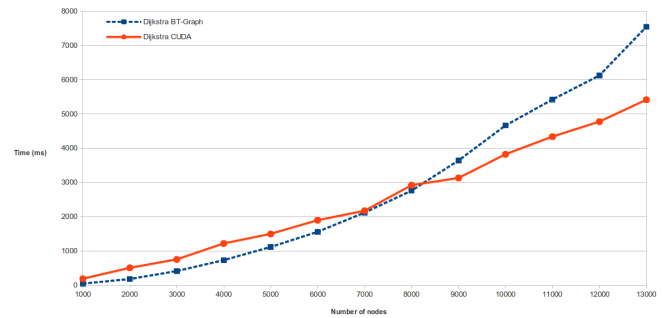

Figure 12. An execution time comparison of serial and parallel Dijkstra's algorithms.

6.6. Experiment 6: Parallelization Dijkstra algorithm for data dissemination

The purpose of this experiment is to compare search speed of traditional Dijkstra for BT-Graph and the corresponding speed of parallel Dijkstra (Dijkstra CUDA). The first step is to construct the weight matrix of BT-Graph model. Next, to obtain the average execution time, serial Dijkstra's algorithm and parallel one are run in turns in 10 times. These experimental results are illustrated in Figure 12.

From Figure 12, CPU can process efficiently for a small a number of vertices because it not waste extra time copying the data between CPU and GPU memories. Nevertheless, when a number of vertices are increased more than 8000, the duration time required for processing can be reduced significantly by using techniques of parallelism based on CUDA.

7. Conclusion

Based on BT-Graph, we have described a new approach in order to model and optimize the dynamic sensor field for LEO satellite connections. The distances between the sub-satellite points and each node of the sensor field is utilized as a key factor to find out the proper gateways for the longest connection time. The experimental results were obtained by applying several appreciate algorithms on BT-DYNSEN model to verify the connectivity of network, determine sensor sets at visiting time, choose set of gateway nodes and find shortest path for data dissemination in DSF. In order to improve the execution time, we implemented the corresponding parallel algorithms with CUDA on GPU, and analyzed its performance. The results show that the parallel algorithm on GPU is considerably superior to the serial one on CPU when BT-Graphs comprising more than 8000 vertices. Thus, our proposed graph-based model helps to increase the amount of time for data communications in long-range sensor field applications using satellite connections.

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