Research Article

Probability Model Based Coverage-Enhancing Algorithm for WSNs of Nodes’ Adjustable Movement Pattern

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Aimed at the defects of current coverage-enhancing algorithms, firstly the probability model is introduced into the coverage-enhancing algorithm of the directional sensor network based on the virtual potential field, and the probability model would have an impact on the location of the centroid. Secondly, to the boundary coverage, through the simulation of typical algorithms which consider the boundary repulsion, it finds that the coverage rates of the corners are much lower than the overall coverage. To add a vertex force on the corner is essential and to some extent would reduce the gap between them. Finally, based on the random deployment of sensor nodes, a novel algorithm is proposed which determines whether the sensor node rotates or moves along fixed direction according to the coverage effect, named as PRMCA (probability model based rotate or move along fixed direction coverage-enhancing algorithm). A set of simulation experiments verify the performance of the proposed algorithm.

1. Introduction

At present, coverage, which is regarded as monitor of service quality, has been an essential problem of the wireless sensor networks (WSNs) [1]. One of these research directions is to use less quantity of sensor nodes or shorter time to achieve the same effect under the premise of ensuring a certain quality of coverage by some algorithm strategies. As the nodes’ deployment reflects the cost and performance of WSN [2, 3], the covering control algorithms can greatly reduce the number of nodes, can also prolong the active time, and naturally can reduce the energy consumption of WSNs [4]. In the WSN coverage research, these aspects need to be considered, including the nodes’ deployment, the nodes’ sensing and communication range, energy efficiency, the algorithm characteristics, and sensor mobility. In different application fields [5, 6], researchers have proposed a lot of coverage mechanisms.

In the traditional research of utilizing the virtual potential field to optimize the quality of WSNs, while calculating the virtual force, some algorithms simply calculate the composition of the force based on the Boolean model so that they ignore the unique attribute of the probability model. For example, Li et al. in [7] proposed the virtual force algorithm (VFA). Tao and Ma in [8] proposed the target involved virtual force algorithm (TIVFA). Tao et al. in [9] proposed the PFCEA for the direction model (potential field based coverage-enhancing algorithm, PFCEA), and so forth. The authors in [10] proposed a coverage configuration algorithm based on the probability detection model (CCAP). The authors in [11] proposed a coverage preservation protocol based on the probability detection model (CPP) that makes working nodes in sensor networks as few as possible when network coverage is guaranteed. However, this protocol configures a network using centred control algorithm which limits the network scale, and at present most of the literatures
have not introduced this probability coverage model into three-dimensional sensor networks. In fact, most practical applied wireless sensor networks are deposited in three-dimensional sensor networks so that it will be more accurate if it is simulated in a three-dimensional space [12, 13]. In [12], Bai et al. proposed and designed a series of connected coverage models in three-dimensional wireless sensor networks with low connectivity and full coverage. In [13], Nazrul Alam and Haas studied a truncated octahedron deployment strategy to monitor the network coverage situation.

In the random deployment, the distribution model is mainly based on the following three assumptions: Poisson’s distribution, normal distribution, and random distribution using software simulation [14–17]. In actual deployment, especially for the harsh environment, sensor nodes are usually laid by the air shedding manner [18]. However, it is inevitable to bring about nodes that are not uniformly distributed in the region. If the sensor nodes achieve the enhancement of coverage only by adjusting the perception angle, the result will be impacted by the nodes’ initial position. The maximization of the coverage cannot be really achieved. In the meantime, the free movement of the sensor nodes is not practical due to ignoring the limitation of movement direction.

According to the methods of calculating the centroid position physically and the characteristics of the probability model, the position of the node centroid based on the probability model can be figured out. The position describes the impact of the centroid on the probability model. On the basis of the above, this paper proposes an algorithm based on coverage effect to determine the sensor to rotate or move along fixed direction which is named as probability model based rotate or move along fixed direction coverage-enhancing algorithm PAMCA. This algorithm can be proved to make up some defects of current algorithms.

2. Coverage Enhancement Issues

2.1. Problem Formulation. The probability model based on the coverage-enhancing problem of directional sensing network composed of the nodes which can rotate and move along fixed direction can be described as follows: When a certain number of sensor nodes are deployed randomly in a specific region, how to calculate the composition force of the centroid in the perception region and determine the sensor to rotate or move along fixed direction according to the coverage effect.

2.2. Analysis and Definitions on the Coverage Enhancement Issue in Directional Sensor Networks. Before research, the following assumptions should be firstly made.

(1) All nodes in WSNs are isomorphic, which means that all the nodes have the same perception radius \( R_s \), the same communication radius \( R_c \), the same perception angle \( \alpha \), the same parameters associated with the physical properties, and so forth. The communication radius is more than or equal to 2 times the perception radius.

(2) The nodes work independently and have the capacity to move towards the specific target location accurately.

(3) The nodes have the sufficient energy to complete the process of movement.

(4) Each node can acquire the information of perception direction and control the direction.

(5) In a rectangular target area \( A \), any two sensor nodes deployed randomly are not in the same location.

The relevant definitions are as follows.

**Definition 1.** Probability Perceptual Model: under the premise that the node \( i \) does not have any neighbour nodes, the perception probability of \( i \) to \( j \) in the monitoring region can be described as follows:

\[
 p_{ij} = \begin{cases} 
 1, & d(i, j) \leq r_1, \\
 e^{-\beta(d(i, j) - r_1)}, & r_1 < d(i, j) \leq r_2, \\
 0, & d(i, j) > r_2, 
\end{cases}
\]

where \( r_1 \) and \( r_2 \) are parameters which are adjustable and related to the sensor physical properties, \( \epsilon \) denotes the natural logarithm, and \( d(i, j) \) represents the distance between node \( i \) and node \( j \).

**Definition 2.** Coverage: the coverage of the WSNs \( W \) deployed in the target region \( A \) is marked as \( C_r(W, A) \). The definition is as follows:

\[
 C_r(W, A) = \frac{\int_A v(p) dS}{V(A)},
\]

where \( V(A) \) is the area of the target region \( A \), for any point \( p \in A \), and

\[
 v(p) = \begin{cases} 
 1, & Cd(p) \geq 1, \\
 0, & \text{otherwise}, 
\end{cases}
\]

\( Cd(p) \) represents the count of the nodes which cover the point \( p \).

**Definition 3.** Coverage Factor: the coverage factor \( \mu \) implies the impact of the network coverage when the node moves once, which is defined as follows:

\[
 \mu = \begin{cases} 
 1, & C'_r(W, A) > C_r(W, A), \\
 0, & C'_r(W, A) \leq C_r(W, A), 
\end{cases}
\]

where \( C'_r(W, A) \) means the network coverage after the node movement.

3. Description and Analysis of the Direction Sensor Based on the Probability Model

Due to the reality, the probability model based on directional sensor has a wider area in application. Sensor energy will decay as the time goes, and the sensing radius will decrease; therefore, how to ensure stable time of the regional coverage...
will be a focus of the future research, which means to add the timeline to the four-dimensional coverage. And we should also pay attention to the value of the coverage probability of the regional point when the regional coverage is studied; the greater the probability is, the longer time the points are covered.

Based on the model and abovementioned definitions, the concept of the centroid is introduced in the probability model. It is reasonable to assume that the mass density is proportional to the size of the probability, so the different probabilities have different effects on the position of the centroid. The centroid position can be calculated by the method in physics.

It is assumed that points have the same mass density if the points have the same distance from the sensor, and the density is proportional to the perception probability, the ratio is set as $k$ (k is a constant); therefore, in a circle that the node is regarded as the center and $2\alpha$ as the central angle, mass is uniformly distributed in the same arc. It means that the centroid position of the arc which has the distance $r$ from the node is in the symmetry axis and has the distance $(r \sin \alpha) / \alpha$ from the node, so it can be deduced further that the centroid position of probability model is based on fan-shaped region, as shown in Figure 1.

It is clear that the centroid position is in the symmetry axis $AB$ and set as $x$-axis. According to the centroid formula of rigid object, it can be drawn that

$$X_c = \left[\int_0^{r_2} (x \sin \alpha \alpha) \cdot 2\alpha \cdot k \cdot x \, dx + \int_{r_1}^{r_2} (x \sin \alpha \alpha) \cdot 2\alpha \cdot k \cdot e^{-\beta(x-r_1)} \cdot x \, dx\right]^{-1} \cdot \left(\int_0^{r_2} 2\alpha \cdot k \cdot x \, dx + \int_{r_1}^{r_2} 2\alpha \cdot k \cdot e^{-\beta(x-r_1)} \cdot x \, dx\right).$$

Simplifying further,

$$X_c = \left(2\beta^3 r_1^3 \sin \alpha - 6\beta^3 \left[r_2^2 e^{\beta(r_1-r_2)} - r_1^2\right] \sin \alpha - 12 \left[\beta r_2 e^{\beta(r_1-r_2)} + e^{\beta(r_1-r_2)} - 1\right] \sin \alpha \right)$$

$$\times \left(3\alpha^3 r_1^4 - 6\alpha \left[\beta^2 r_2 e^{\beta(r_1-r_2)} + \beta e^{\beta(r_1-r_2)} - \beta^2 r_1 - \beta\right]\right)^{-1}.$$

where $r_1 = AC$, $r_2 = AD$, $\alpha$, $\beta$ are the positive constants and related to the sensor physical properties, and the centroid position can be calculated directly in the probability model based on fan-shaped perception region. It is in the symmetry axis and has the distance $X_c$ from the center.

Obviously, when $r_1 = r_2$, the centroid is in

$$X_c = \frac{2R \sin \alpha}{3\alpha}.$$

It can be concluded that the uniform sensing region is a special case of the probability perception model. When $\alpha = \pi$, the directional sensor becomes the omnidirectional perception model and the centroid is in the center of the circle.

4. The Study of the Visual Force

When the distance between the sensor nodes is less than $2R_S$, it is possible that the perception region is overlapped, which provides the possibility of repulsion. It is reasonable to assume that if several perception areas are overlapped, its probability should be the maximum of each probability. In addition, according to the probability perception model, it can be concluded that when the probability is less than a certain value, the effect of the node can be ignored. The threshold can be determined by the sensor parameters of $r_1$, $r_2$, and $\beta$. In this paper the possibility threshold is $e^{-\beta(r_1-r_2)}$ and the perception radius can be considered as $R_S = r_2$.

Under the impact of the repulsion, the nodes spread as far as possible in order to ensure the maximum area covered. It is well known that the condition that the distance between sensor nodes is less than $2R_S$ is a necessary condition of the existence of overlapped area instead of sufficient condition. Therefore, it is unreasonable to only rely on the distance between the nodes to judge whether the repulsion exists. Reference [19] points out that there are 11 overlapping situations for perception areas, as shown in Figure 2.

When the distance of the two nodes is less than $2R_S$, there will be no repulsion between the nodes because the overlapping regions are not existed. Therefore, it will not consume energy caused by the action of nodes; in this case the energy can be saved.

Reference [20] provides another judgment for the existence of perception overlapping area: while the perception model is simplified as a triangle, according to the condition of whether the two triangles have intersecting edge, the existence of the overlapping area can be judged. The method is called the straddle experiment. As is shown in Figure 3, there must be at least an intersecting edge for the two triangles with the overlapping regions; in Figure 3, $S_1a_1$ and $S_2a_2$ intersect, and the following conditions must be met:

$$[(a_2 - S_1) \times (a_1 - S_1)] \times [(a_1 - S_1) \times (S_2 - S_1)] \geq 0, \quad (8)$$

$$[(S_1 - a_2) \times (S_2 - a_2)] \times [(S_2 - a_2) \times (a_1 - a_2)] \geq 0. \quad (9)$$
If the relationship between the nodes meet conditions in (8) and (9), there will exist repulsion between nodes, and it will be defined as

$$F_{\text{rep}}(i, j) = \begin{cases} K_{\text{rep}} \left( \frac{i - j}{d(i, j)} \right), & SD_i \cap SD_j \neq \emptyset, \\ 0, & SD_i \cap SD_j = \emptyset, \end{cases}$$

(10)

where $K_{\text{rep}}$ is repulsion coefficient, a positive constant. Therefore, the centroid $n_i$ is the vector sum of the repulsion of adjacent $K$ centroid points. The definition of the force mode $\vec{F}$ is

$$F_{\text{rep}}(i) = \begin{cases} \sum_{j=1, j \neq i}^{K} F_{\text{rep}}(i, j), & K \geq 1, \\ 0, & K = 0. \end{cases}$$

(11)

In the random deployment, it is not easy to realize that the nodes are distributed uniformly. The randomness of the nodes’ position is great, so it is possible that many points are centralized. In the calculation of the virtual joint force, the boundary problem should be taken into account, which is not considered in most of the papers, or just one boundary direction, as shown in Figure 4.

In the simulation of the proposed algorithm in this paper, it takes four same squares in the corner, whose area is 1/9 of the total area but other algorithms only consider the repulsion nearest to boundary nodes. The number of adjustment is set 100.

As is shown in the simulation experiment results of the Table 1, it is not enough to only consider the repulsion within a certain distance to the node $D$ in the corner. The attraction of the uncovered vertex $B$ should also be taken into consideration, namely, the attraction of the vertex to the sensor.

This paper assumes that the force of the nearest boundary points and four vertexes of the rectangular frame to the sensor node exist. Obviously, in the region that needs to be covered, the force of all directions should be considered simultaneously if all the vertexes are in the communication range of the sensor node.

As is shown in Figure 4, $C$ is the centroid point, $D$ is a boundary point within a certain distance to sensor node, and $B$ is a border vertex; in this case the force of $D$ and $B$ on $C$ must be considered.
The definition of the boundary force is as follow:

\[
F_{\text{bor}}(i) = \begin{cases} 
\frac{k_{\text{bor}}}{d(i,q)^2}(\frac{x_i - q}{d(i,q)}), & d(i,q) \leq \sqrt{3}r_s, \\
0, & d(i,q) > \sqrt{3}r_s,
\end{cases}
\]

(12)

where \(k_{\text{bor}}\) is ratio coefficient, a constant. \(\sqrt{3}r_s\) is a threshold distance, which is mentioned above as the value of a certain distance, when the distance is below this value, the force of the boundary point and vertex starts to take effect on nodes.

For the direction of the force, because the node is close to the boundary, the coverage area may be reduced due to being out of the region. Therefore, it is reasonable to assume that there exists repulsive force between the boundary points and the sensor node. However, if the vertexes are not covered, there will exist a blind zone in the region. In this case, if all the force is repulsion, the blind zone will not be erased because the sensor node cannot get close to the corner. So in this paper, a new solution will be introduced: if the vertex is not covered, the attractive force will take effect while if the vertex has been covered, the repulsive force will make the overlapping area eliminated quickly. The attractive force is set by a negative value.

To sum up, the resultant force of the sensor centroid \(i\) can be expressed as

\[
F(i) = F_{\text{rep}}(i) + F_{\text{bor}}(i).
\]

(13)

5. Descriptions of Node Movements

Now, in most papers, when considering the forces of the sensor nodes, the researchers only consider the rotation or the omnidirectional movement of the nodes, while the omnidirectional movement is impractical and the rotation of the node costs too much energy when dispersing without a high speed. For instance, when considering the rotation merely, only when the force has a tangential component, the node can rotate along this direction. As Figure 5(a) shows, though there is an overlapping region, it has no forces in the tangential direction. Therefore, the node can’t rotate; even if it rotates, it also needs to rotate for more times to make the overlapping region disappear. As Figure 5(b) shows, when considering the movement merely, it only needs to rotate for a small angle to make the overlapping region disappear. But in this condition, the movement will certainly cause the unnecessary energy cost. Hence, a method can be introduced that movement and rotation are considered synthetically to avoid consumption of the node’s energy.

In the movement procedure of nodes, their moving directions are not arbitrary. Such as cameras, they can only move forward and backward by focusing or modulate the viewpoint by rotating. In conclusion, most moving directions of sensor nodes are limited. Therefore, this paper assumes that the node only rotates or moves along with the sensor’s direction.

In physics, acceleration has been defined. It is a vector that measures the speed changing in unit time. It reflects the moving trend of the object. This conception can be applied in coverage researches. According to the analysis above, a criterion must be introduced to measure the moving change of objects. We can use the coverage improvement (influence factors of coverage) as a measurement and call it acceleration.

Objects’ movement can be judged by the influence factors of coverage. We consider that the influence factors of coverage (acceleration) in this direction are bigger than the other direction; the node will move towards this direction. Otherwise, the node will move towards the other direction. In particular, we define that if the accelerations in two directions are equal, the sensor will rotate. It shows as Figure 1.

That, the node can only move along with the center line (\(AB\) direction or the opposite direction) or the tangential direction of the fan-shape. Therefore, we can determine the final moving direction of the node according to the influence of resultant force \(\vec{F}\) and \(\theta\).

\[
\vec{F} \cdot (\cos \theta, \sin \theta) \geq 0,
\]

(14)

\[
(\cos \theta, \sin \theta) \times \vec{F} \geq 0.
\]

(15)

When (14) was founded, the angle between resultant force \(\vec{F}\) and \(\vec{AB}\) is less than \(\pi/2\) and the node moves along with \(\vec{AB}\). Otherwise, it will move towards the negative direction of \(\vec{AB}\). When (15) was founded, \(\vec{AB}\) rotates along with the tangential direction and the angle \(\theta\) increases. On the contrary, \(\theta\) decreases. More attentions need to be paid to the boundary value. When the node crosses the region boundary in its movement process, the node must be on the boundary. In the rotation process, when \(\theta\) is greater than \(\pi\), \(2 \times \pi\) needs to be subtracted from \(\theta\). Similarly, when \(\theta\) is less than \(-\pi\), \(\theta\) needs to add \(2 \times \pi\).

6. Algorithm Description

A coverage-enhancing algorithm is introduced as follows according to the virtual forces to nodes. This algorithm is a distributed one, which runs simultaneously in each sensor. Taking node \(i\) for example, its formal description is shown as in Algorithm 1.
Algorithm 1

//initialization
(1) \( t \leftarrow 0; \)
(2) Set relevant parameters and coefficients;
(3) Set the max cycle-index \( t_{\text{max}} \);
While \( (t < t_{\text{max}}) \) do
{
(4) Renew the neighbor nodes set of node \( i \);
(5) Read the current coordinate position of node \( i \);
(6) Read all neighbor nodes of node \( i \);
(7) Calculate resultant force to node \( i \) according to formulas (8)–(13);
(8) Calculate the target position of node \( i \) at time \( t + 1 \)
    according to formulas (14) and (15);
(9) Calculate the coverage rate of this movement and rotation
    according to formula (2);
(10) Determine the mode of movement according to the
    coverage rate. Calculate influence factors \( \mu \). And determine
    the final moving direction of the node.
(11) If \( \mu = 1 \), move node \( i \) according to the mode and direction
    in step (10), or keep the current position of node unchanged.
(12) \( t \leftarrow t + 1; \)
}
(13) End.

Figure 5: Two special situations in the WSNs.

7. Algorithm Simulations and Performance Analysis

This simulation experiment uses VS2010 to develop and realize a probability model based coverage-enhancing algorithm for WSN of nodes’ adjustable movement pattern. And it uses Matlab 2009 to achieve many simulation experiments, through which the effectiveness and advantages of this algorithm have been proved. To simplify these experiments, it is assumed that all assumptions in Section 2.2 are met.

7.1. Experiment Parameters’ Setting. In the simulation, it sets a 300 m \( \times \) 300 m sized square target region. Some nodes are distributed in this region randomly. Repulsion coefficient among nodes is defined as \( k_{\text{rep}} = 100 \). Gravitation coefficient among nodes can be described as \( k_{\text{att}} = 0.025 \). Gravitation index \( \partial = 1 \). Repulsion coefficient among boundaries is \( k_{\text{bor}} = 1 \). In the probability model, we can set the parameter \( \beta = 1 \).

7.2. Algorithm Convergence Analysis. To discuss the convergence of PRMCA, some experiments are carried out using four kinds of different network nodes’ number. We produce ten scenarios randomly in terms with different topological
constructs and network nodes’ number, meanwhile calculating the average value of the algorithm’s convergence. Experiment data are showed in Table 2. Other experiment parameters are defined, respectively, as \( R_1 = 20 \), \( R_2 = 40 \) m, and \( \alpha = \pi/6 \).

Obviously, it can be concluded as follows: the convergence of PRMCA (regulating times) will not have an obvious change as the increase of the number of sensor nodes. The range of variation is usually between 20 to 30. Therefore, PRMCA proposed in this paper has a satisfactory convergence. It can accomplish the coverage-enhancing process in a short limited time.

![Figure 7: Coverage optimization under PRMCA.](image)

Table 2: Experiment data convergence analysis.

| Number of nodes/N | Original coverage rate/% | Final coverage rate/% | Cycle-index/t |
|-------------------|--------------------------|----------------------|--------------|
| 50                | 30.4891                  | 41.6233              | 22.3         |
| 100               | 50.3642                  | 67.2216              | 24.7         |
| 150               | 61.6025                  | 85.3278              | 28.2         |
| 200               | 77.5234                  | 98.127               | 27.6         |

7.3. Case Study. In this section, PRMCA will be analyzed through cases:

\[
N \geq \frac{\ln(1 - P)}{\ln(S - \alpha R_2^2) - \ln S},
\]

where \( P \) can be defined as the coverage rate you expected and \( S \) can be described as the total region area. According to the experiment parameters, the number of sensor’s nodes can be estimated if the network coverage rate reaches \( P \) as expected. Such as these cases, if \( P = 70\% \), then \( N = 129 \); if then \( P = 80\% \), then \( N = 172 \); if \( P = 90\% \), then \( N = 246 \). In these simulations, \( N = 100 \), \( N = 150 \), and \( N = 200 \).

For the example above, when \( N = 150 \), the movement process can be shown in Figure 7.

In Figure 7, nodes tend to uniformly distributive gradually in the influence of the virtual force. The proposed algorithm can achieve the enhancement by the adjustment which can eliminate the overlapping and the blind regions. After 30-step adjustments, the coverage rate increases from 62.375% to 78.496% with the enhancement of 16.121%.

Figure 6 shows that due to the influence of the coverage factor, the coverage rate is constantly improving. When it reaches around 15 times, coverage rate tends to be stable,
the effect on the coverage rate enhancement is not quite remarkable. The movement of the node would make energy consumption, so it will save energy without moving any longer.

7.4. Simulation Analysis. The node in each movement must choose the path containing the maximum coverage rate, because the algorithm proposed in this paper is based on the acceleration which determines the movement mode, [9] for the two-dimensional space of the wireless sensor network provides an enhancing algorithm PFCEA based on the virtual force. It introduces the coverage factor into the algorithm and names the improved algorithm as PFCCEA. By comparison, as Figure 8 shows, PRMCA converges much faster than PFCEA. In addition, node in PFCEA needs to be randomly and uniformly distributed, so PRMCA obtains better performance than PFCEA. But at the same time, the initial location of the node based on PFCEA has the greater impact.

This section is also to see the influence by changing the parameters of the proposed algorithm, which includes the number of nodes \( N \), perception angle \( \alpha \), and sensing radius \( R_S \). Performance analysis and comparison between PFCEA and PRMCA are demonstrated by Figure 9.

From the curve of Figure 9, it can be seen that when \( N \) and \( R_S \) are fixed, the smaller the value of \( \alpha \), the smaller coverage rate will be. With the increasing of \( \alpha \), the increasing of coverage rate \( \Delta p \) tends to be higher, and when \( \alpha = \pi/3 \), network coverage increase can reach 20.7076\%. After that,
Δp drops. This is because with the increasing of the perceived angle, the probability that covers blind spot is reduced greatly, which will undoubtedly sacrifice the PRMCA performance.

Obviously, from Figures 9(b) and 9(c), it can be seen that the sensing radius $R_S$ and the node number $N$ have the similar effects on the performance of PRMCA algorithm. When the sensing angle $α$ is certain, the value of sensing radius $R_S$ or node number $N$ is larger, the possibility that blind spot exists is smaller, and the performance of PRMCA becomes not obvious.

In addition, through the comparison, parameters such as nodes scale, sensing radius, and sensing angle have similar effect on the performances of PFCCEA and PRMCA. However, due to the limitations of PFCCEA that PFCCEA does not consider the boundary effects and PFCCEA requires uniform deployment of the nodes, it is bound to be limited by the initial coverage. At the same time, it will limit the improvement in the network coverage performance. By contrast, PRMCA is dominant.

7.5. Analysis on the Boundary. As shown in Figure 10, through the introduction of the force in corner and vertices, to a certain extent, the algorithm can reduce the gap between the overall coverage and boundary coverage. Owing to omitting the local effect of the nodes, the boundary coverage is lower than the overall coverage.

8. Conclusions

In this paper, the probability model based on virtual potential field is introduced into the coverage-enhancing algorithm, whose probability is reflected by the position of “centroid” and transforms the sensor network coverage enhancement problem into a problem that the center of mass gradually spreads by the virtual force, and the blind region and the overlap zone are eliminated.

It is difficult for the nodes to achieve the uniform distribution in the actual application, especially in the dangerous area. At the same time, the moving direction of the node was restricted, such as camera. In this way, the initial node will ultimately affect the entire regional coverage. The proposed PRMCA can let the nodes select a movement form according to the “acceleration,” which can let the node achieve the convergence faster and save the energy of nodes.

Because the algorithm is an optimal algorithm, so the experimental data can be seen that the corner coverage must be smaller than the overall coverage. By considering the gravity in boundary vertices, it improves the corner coverage, but it is still lower than the overall coverage.

Finally, through a series of simulation experiments, it verifies the convergence and effectiveness of PRMCA, and describes several important parameters of PRMCA that influence performance. In this paper, it effectively improves the WSNs coverage rate, but when one point is covered by many sensors, the integrated calculation of this point is not considered, and it would be taken into account in the further research in the four-dimensional space.

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References

[1] S. Meguerdichian, F. Koushanfar, M. Potkonjak, and M. B. Srivastava, “Coverage problems in wireless ad-hoc sensor networks,” in Proceedings of the 20th Annual Joint Conference of the IEEE Computer and Communications Societies (INFOCOM ’01), vol. 3, pp. 1380–1387, IEEE Press, New York, NY, USA, April 2001.

[2] J. Huang, L. Sun, R. Wang, and H. Huang, “Improved virtual potential field algorithm based on probability model in three-dimensional directional sensor networks,” International Journal of Distributed Sensor Networks, vol. 21, no. 6, pp. 45–53, 2012.

[3] E. Onur, C. Ersoy, and H. Delic, “Analysis of target detection probability in randomly deployed sensor networks,” IEEE Communications Letters, vol. 11, no. 10, pp. 778–780, 2007.

[4] Y. Wu, M. Li, Z. Cai, and E. Zhu, “A distributed algorithm to approximate node-weighted minimum $α$-connected ($θ,k$)-coverage in dense sensor networks,” in Proceedings of the International Frontiers of Algorithmic Workshop, vol. 5059, pp. 221–232, Springer, Berlin, Germany, 2008.

[5] M. Cardei and J. Wu, “Energy-efficient coverage problems in wireless ad-hoc sensor networks,” Computer Communications, vol. 29, no. 4, pp. 413–420, 2006.
[6] Y. Zou and K. Chakrabarty, “Sensor deployment and target localization in distributed sensor networks,” ACM Transactions on Embedded Computing Systems, vol. 3, no. 1, pp. 61–91, 2004.

[7] S. J. Li, C. F. Xu, Z. H. Wu, and Y. H. Pan, “Optimal deployment and protection strategy in sensor network for target tracking,” Acta Electronica Sinica, vol. 34, no. 1, pp. 71–76, 2006 (Chinese).

[8] D. Tao and H. Ma, “Coverage-enhancing algorithm for directional sensor networks,” in Proceedings of the of the 2nd International Conference on Mobile Ad-Hoc and Sensor Networks, I. Stojmenovic and J. N. Cao, Eds., pp. 256–267, Springer, Berlin, Germany, 2006.

[9] D. Tao, H. D. Ma, and L. Liu, “A virtual potential field based coverage-enhancing algorithm for directional sensor networks,” Journal of Software, vol. 18, no. 5, pp. 1152–1163, 2007 (Chinese).

[10] D. Zhang, M. Xu, Y. Chen, and S. Wang, “Probabilistic coverage configuration for wireless sensor networks,” in Proceedings of the International Conference on Wireless Communications, Networking and Mobile Computing (WiCOM ’06), pp. 1–4, September 2006.

[11] J. P. Sheu and H. F. Lin, “Probabilistic coverage preserving protocol with energy efficiency in wireless sensor networks,” in Proceedings of the IEEE Wireless Communications and Networking Conference (WCNC ’07), pp. 2633–2638, March 2007.

[12] X. Bai, C. Zhang, D. Xuan, and W. Jia, “Full-coverage and k-connectivity (k = 14, 6) three dimensional networks,” in Proceedings of the 28th Conference on Computer Communications (IEEE INFOCOM ’09), pp. 388–396, April 2009.

[13] S. M. Nazrul Alam and Z. J. Haas, “Coverage and connectivity in three-dimensional networks,” in Proceedings of the 12th Annual International Conference on Mobile Computing and Networking (MOBICOM ’06), pp. 346–357, Los Angeles, Calif, USA, September 2006.

[14] B. Liu and D. Towsley, “A study of the coverage of large-scale sensor networks,” in Proceedings of the IEEE International Conference on Mobile Ad-Hoc and Sensor Systems, pp. 475–483, October 2004.

[15] M. Cardei and D. Z. Du, “Improving wireless sensor network lifetime through power aware organization,” Wireless Networks, vol. 11, no. 3, pp. 333–340, 2005.

[16] D. Brinza and A. Zelikovsky, “DEEPS: deterministic energy-efficient protocol for sensor networks,” in Proceedings of the 7th ACIS International Conference on Software Engineering, Artificial Intelligence, Networking, and Parallel/Distributed Computing (SNPD ’06), pp. 261–266, June 2006.

[17] D. Tian and N. D. Georganas, “A coverage-preserving node scheduling scheme for large wireless sensor networks,” in Proceedings of the 1st ACM International Workshop on Wireless Sensor Networks and Applications, pp. 32–41, Atlanta, Ga, USA, September 2002.

[18] J. Wang, C. Niu, and R. Shen, “Priority-based target coverage in directional sensor networks using a genetic algorithm,” Computers and Mathematics with Applications, vol. 57, no. 11-12, pp. 1915–1922, 2009.

[19] D. Pescaru, C. Istin, D. Curic, and A. Doboli, “Energy saving strategy for video-based wireless sensor networks under field coverage preservation,” in Proceedings of the IEEE International Conference on Automation, Quality and Testing, Robotics (AQTR ’08), pp. 289–294, May 2008.

[20] J. J. Huang, L. J. Sun, R. C. Wang, and H. P. Huang, “Virtual potential field and covering factor based coverage-enhancing algorithm for three-dimensional wireless sensor networks,” Journal on Communications, vol. 31, no. 9, pp. 16–21, 2010.
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