

Proposal of a new Maximum Lifetime Communication Model of Wireless Sensor Network

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Abstract—One new maximum lifetime communication model of wireless sensor network is proposed in this thesis and the model is based on conditions of base station with multiple sources and multiple links and the data generated from the source node can be transmitted to several base stations through multiple links. In this chapter, energy limitation and data stream conservation principles of a wireless sensor network are used to establish a linear planning model, and attaining the maximum lifetime of the network is the optimal objective to establish a model with multiple restrictions restricted by energy and bandwidth. The basic element influencing the lifetime of a wireless sensor network is the data forwarding rate. The lifetime of nodes will become longer with the decrease of node data. In this thesis, the data size of network nodes is taken as sub-optimization objective and the distributed heuristic algorithm will be used to solve the linear planning model. In the end, a simulation experiment is conducted to verify the performance of the communication model and indicate that the proposed model can effectively increase network lifetime.

Index Terms—maximum lifetime, communication model, wireless sensor network.

I. INTRODUCTION

As shown in Fig. 1, the wireless sensor network consists of a sensor node, sink and manager node. A large number of sensor nodes are randomly distributed inside or near the monitoring area after being dripped and all nodes constitute the network through self-organization. The sensor node will monitor objects and data monitored in the initial processing will be transmitted by a multi-strip relay-operated method according to specific routing protocol. In the transmitting process, monitored data will be processed by several nodes and then they are routed to the aggregation nodes[1]. Then the data will be transmitted to manager nodes through a satellite, internet and mobile communication network. The end user will carry out configuration and management on the sensor network, issue a monitoring task and collect monitoring data.

In the multiple base station network of a wireless sensor, the network lifetime is increased to some degree when compared with the base station network. The network forms multiple groups in multiple base stations among which there are several similar route models to establish LEACH algorithm so as to form several clusters. The base

station will be used to substitute cluster head in LEACH algorithm and every base station maintains surrounding node communication links. Furthermore, a hierarchical tree-form communication structure CWSTP [2-3] is established for every base station and surrounding nodes in the network and each node only needs to communicate with the nearest base station, which can relieve data transmission pressure of the surrounding nodes in the base station, thus increasing network lifetime. As for nodes[4-5], one communication algorithm will be selected from multiple base stations, which is relatively similar to any cast communication technology in the TCP/IP protocol. Tony proposes one algorithm (MD-MLR) to realize maximum lifetime and minimum delay of wireless sensor network by any cast technology, and it mainly realizes the maximum lifetime of the network in data delay restrictions [6]. However, the communication model with multiple base stations is still insufficient in relieving communication pressure of surrounding nodes of the base station. In fact, the network is divided into several task areas according to distribution conditions of the base station and every base station is responsible for information collection conditions in one task area. To a certain degree, it is the collection of several models of a base station, which only shortens the node scale of base station maintenance in a single base station model.

One new maximum lifetime communication model of wireless sensor network is proposed in this thesis, and the model is based on conditions of a base station with multiple sources and multiple links. The data generated from the source node can be transmitted to several base stations through multiple links[7]. In this chapter, energy limitation and data stream conservation principles of a wireless sensor network are used to establish a linear planning model and maximum lifetime of network is the optimum objective to establish a model with multiple restrictions restricted by energy and bandwidth. The basic element influencing the lifetime of a wireless sensor network is the data forwarding rate[8]. The lifetime of nodes will become longer with the decrease of node data. In this thesis, the data size of network nodes is taken as sub-optimization objective and the distributed heuristic algorithm will be used to resolve the linear planning model. In the end, a simulation experiment is conducted to verify the performance of the communication model and indicate that the proposed model can effectively increase network lifetime, as shown in Figure 1.

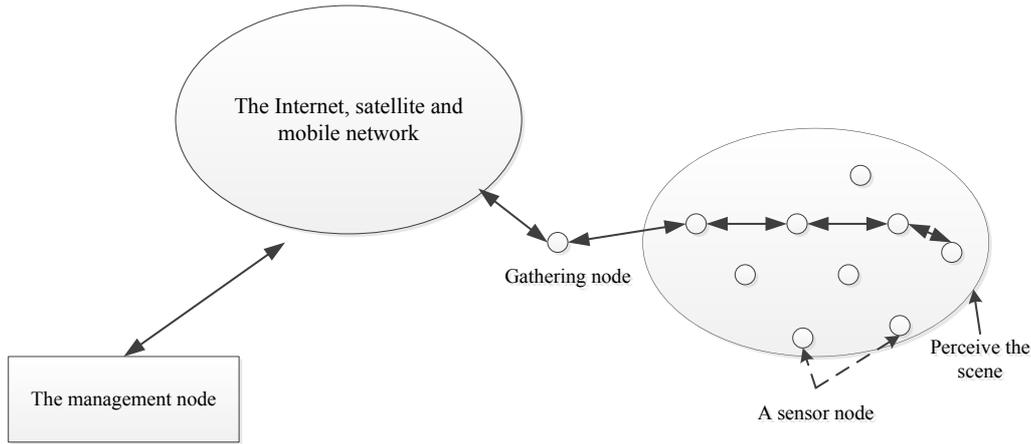


Figure 1. The architecture of wireless sensor network

II. MLCM SYSTEM MODEL

A. Network model

The wireless sensor network consists of a large number of sensor nodes which are densely distributed in a monitoring area for operation without engineering treatment or propositioning. Therefore, necessary sensor nodes will have self-organization characteristics. At the same time, multiple routes are necessary for transmission of data for the sensor nodes are embedded systems and the processing capacity, storage capacity and communication capacity are relatively weak. Additionally, apart from local information collection and data processing, nodes must store, manage and integrate data transmitted by other nodes. Furthermore, they must coordinate with other nodes and transmit original data to an aggregation node after processing[9-10]. As the interface of the sensor network and external network, the aggregation node can realize the communication between manager nodes and the wireless sensor network through network protocol. It will forward data collected from the sensor nodes to the external network and issue commands from the manager node to the sensor network at the same time, as shown in Figure 2.

The wireless sensor network can be abstracted as one undirected graph among $G(V,A)$ where V represents the set of sensor nodes and base stations in the network and A represents the set links. In the case that the set of sensor nodes in network is S , the node number is n , the set of base station is s and the number of base stations is m , then $V=S \cup \bar{S}$. In the case that the set of adjacent node of $V_i \in V$ is S_i , then V_i can have direct communication in any nodes inside S_i . The communication set of node V_i is X , $X \subseteq S_i$, and X consists of nodes between node V_i and base stations and the communication link of node V_i is (V_i, X) . The link set can be expressed as: $A = \{(V_i, X) | V_i \in V, X \subseteq S_i\}$. The source node V_k will perceive and generate data which will finally arrive at the base station node S_d through one or several frontier sets (V_i, V_j) provided that the data size surpasses the frontier set (V_i, V_j) in unit time in that process is f_{V_i, V_j}^a . In the case that the primary energy of all sensor nodes is E , then the energy of node V_i is E_i .

B. Fundamental assumption

The assumption is that the sensor network is of the following characteristics:

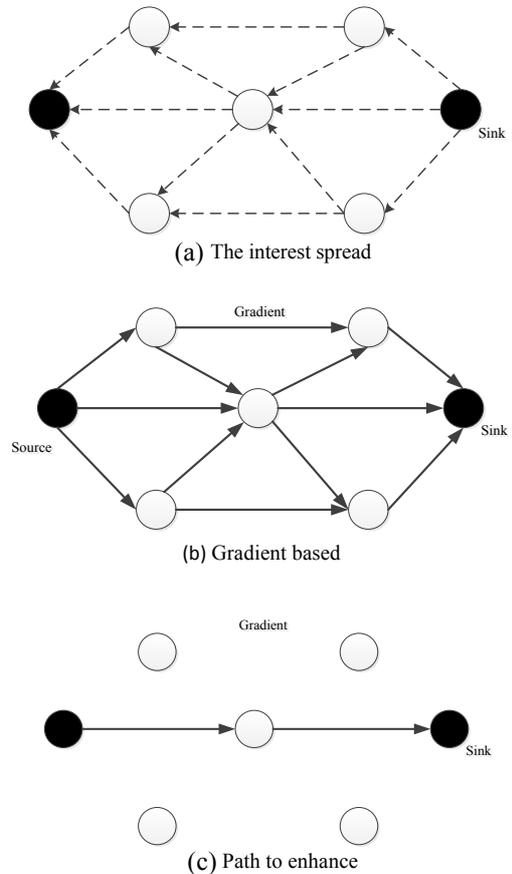


Figure 2. Network type

- (1) The nodes in the network are uniformly distributed and the sensor nodes and the base station are stationary.
- (2) All sensor nodes have the same initial energy and once the energy of a certain node is used up, the node will stop working.
- (3) All nodes can transmit and receive data and partial nodes selected to periodically induce the data are called source nodes[11].
- (4) All nodes have the same maximum transmission range d_{max} and the nodes with a distance from the base station is more than d_{max} the data is transmitted to the base station through the multi-hop routing mode.
- (5) Networks are interconnected initially.

C. Energy consumption model

The energy consumption in the node of the wireless sensor network includes three parts, which are respectively the processor module, sensor module and wireless communication module, of which the wireless communication consumes the most energy of the sensor node. The energy consumed by the wireless communication module in the sensor node is researched in this thesis, and the algorithm adopts the energy model used in Literature [12-13].

Each sensor node $V_i \in S$ continuously or periodically reports the data collected to the base station and the energy consumption of V_i transmitting unit data to the neighbor node is:

$$e_{ij} = \alpha + \beta \cdot Range_i^m \quad (1)$$

Assuming the coordinates of the node and communication object K respectively are (X_i, Y_i) and (X_j, Y_j) , then the communication radius is:

$$g_i = \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} \quad (2)$$

In the case that the node does not have the positioning function and the node transmitting power is non-adjustable, the node generally broadcasts to the surrounding nodes with a fixed transmitting radius. Assuming the data generation rate of node V_i is g_i , then:

$$e_t(i, j) = e_{ij} \cdot \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} \quad (3)$$

Assuming the link bandwidth is R and it is allowed that the maximum data transmission amount of node V_i and the adjacent node V_j in unit time is R_{ij} , and then it is obvious that $f_{V_i V_j}^{V_i \bar{S}_d} \leq R$. The energy consumption of communication between node V_i and the adjacent node V_j is $e(i, j)$.

$$e_r(i) = \rho \cdot \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} \quad (4)$$

The energy consumption of node V_i receiving the data of adjacent node V_j is $e_r(i)$ and $e_r(i)$ is unrelated to the distance between them, where, ρ is the fixed constant. Other possible energy consumption of the node include e_g representing the energy consumption of the node generating unit data, e_{idle} the energy consumption of node in an idle state and e_{listen} the energy consumption of node continuously listening to the network state so as to promptly awaken the node. These three sources of energy consumptions are the same for all nodes and the energy consumption value is basically fixed, and given $e_{other} = e_g + e_{idle} + e_{listen}$, the energy consumption of the node V_j within unit time is

$$e(i, j) = e_{ij} \cdot \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \rho \cdot \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + e_{other} \quad (5)$$

D. Communication model

The maximum lifetime communication model of the wireless sensor network proposed in this thesis is based on a multi-source, multi-link and multi-base station condition. The data generated by the source node can be transmitted to different base station nodes through different links[14].

The source node generates data, the intermediate node provides data forwarding service, the data continuously converges to the base station node from the edge node in the network and the source node induces the data generated and finally arrives at the base station node through single-hop or multi-hop forwarding. Data generated by nodes of the same source may be routed to different base station nodes through different links and it is of very strong dependency[15]. Different source nodes may conduct data fusion processing in the process of data forwarding. On account of the base station node requiring data from the same source for the fusion processing, multiple base station nodes form a tree structure and the root node is connected with the virtual link through each base station, of which one base station node is taken as the root node and it conducts redundancy removal to compress all data and then sends it to the terminal receiver.

III. MLCM COMMUNICATION MODEL

A. Maximum lifetime model

The wireless sensor network is the energy restriction network and the energy consumption of the node will not exceed the residual energy. Based on Equation (7), when the data forwarding rate of the node is f , the lifetime of a node V_i is

$$T_i(f) = \frac{E_i}{p_r(i) + p_t(i, j) + e_{other}} = \frac{E_i}{e_{ij} \cdot \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \rho \cdot \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + e_{other}} \quad (6)$$

The nodes of the wireless sensor network under working conditions based on the effect of node on the data stream can be divided into source node and intermediate node. The source node induces, generates and receives data and then forwards the data to the next adjacent node. The intermediate node does not generate data, but only receives data and then forwards to the next adjacent node. The data stream conditions of the two nodes are:

$$\sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i = \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} \quad (7)$$

Where,

$$\lambda_i = \begin{cases} 1, V_i \neq \bar{S}_d \\ -1, V_i = \bar{S}_d \end{cases} \quad (8)$$

Based on Equation (8), the lifetime model of network node is:

$$T_i(f) = \frac{E_i}{e_{ij} \left(\sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i \right) + \rho \cdot \sum_{V_i, V_j \in S, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + e_{other}} \quad (9)$$

The lifetime $T_{sys}(f)$ of the network at a certain data forwarding rate f is the minimum lifetime of all nodes.

$$T_{sys}(f) = \min_{i \in S} T_i(f) \quad (10)$$

The objective of the energy-efficient network communication model is to maximize the minimum lifetime of the node and prolong the network lifetime as much as possible to make the energy consumption of a single node balanced with the whole network[16]. The purpose of research on the maximum lifetime communication model of wireless sensor network is to

maximize the network lifetime. The maximum lifetime communication model established in this chapter is:

$$\maximize \quad \min_{i \in S} T_i(f) \quad (11)$$

$$\text{subject to} \quad \sum_{V_i, V_j \in \bar{S}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i = \sum_{V_i, V_j \in \bar{S}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d}, \forall V_i, V_j \in V, \forall \bar{S}_d \in \bar{S} \quad (12)$$

$$T_i(f) (e_{ij} \cdot \sum_{V_i, V_j \in \bar{S}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i + \rho \cdot \sum_{V_i, V_j \in \bar{S}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + e_{other}) \leq E_i \quad (13)$$

$$\sum_{\bar{S}_d \in \bar{S}} f_{V_i V_j}^{V_i \bar{S}_d} \leq R_{ij}, \forall V_i, V_j \in X, X \subseteq S_i \quad (14)$$

$$f_{V_i V_j}^{V_i \bar{S}_d} \geq 0, \lambda_i = 0 \text{ or } 1, g_i \geq 0, \forall V_i, V_j \in X, X \subseteq S_i \quad (15)$$

The Condition 1 in Equation (12) shows that the quantity of data transmitted by node Vi to the adjacent node is equal to the sum of data generated by the data induction and the quantity of data received from the adjacent node, and the data loss in transmission is not considered in this chapter. Condition 2 shows that the total energy consumption of node I within time of $T_{sys}(f)$ will not exceed the set value E_i of the node energy. Condition 3 shows that the communication rate between the node and the adjacent node is less than the network bandwidth. The condition of the node not participating in the network communication is not considered in this thesis, such as the conditions of the dormant node, a node without a monitoring area and the node that has accidentally gone dormant in deployment[17].

B. Data stream model

The base station set $\bar{S} = \{\bar{S}_1, \bar{S}_2, \dots, \bar{S}_m\}$ can form $2^m - 1$ non-void subsets. In the case of assuming $m=2$ and $\bar{S} = \{\bar{S}_1, \bar{S}_2\}$, the subsets possibly formed are $\{\bar{S}_1\}$, $\{\bar{S}_2\}$ and $\{\bar{S}_1, \bar{S}_2\}$. In each process of the source node transmitting data, the data will be transmitted to a certain subset of the base station. In the adjacent communication link $\{V_i, X\}$ of the source node, the set X is not void with $X \subseteq S_i$ and $X = \{X_1, X_2, \dots, X_k\}$; the data of mode Vi is routed to one base station as the least through k links and assuming the node Vi passes through the adjacent node X1 and transmits to the subset P1 of the base station, and transmits the subset P1 of the base station through node X2; that is, the set transmitted to the base station subset through the adjacent communication link set X is

$$P = \{P_1, P_2, \dots, P_k\} \text{ and } P \in \bar{S} \quad (16)$$

The subset of node Vi transmitted to the base station through the adjacent node V_j is P_k with $P_k \notin \Phi$ and $P_k = \{P_1, P_2, \dots, P_k\}$; assuming that the data forwarded by node Vi through the adjacent node V_j is received by P_k base station subset at the average data rate of $F_{V_i, V_j}^{V_i}$ and total amount of data received by P_k base station is $F_{V_i, V_j}^{V_i} \{P_1, P_2, \dots, P_k\}$.

In the communication model, $F_{V_i, V_j}^{V_i, S_d}$ represents the quantity of data passing through the edge set $\{V_i, V_j\}$ within a unit of time and based on the data stream model, the relation between $F_{V_i, V_j}^{V_i, S_d}$ and the average data rate $F_{V_i, V_j}^{V_i}$ arriving at the base station set is:

$$\sum_{V_j \in X, X \subseteq S_i} f_{V_i V_j}^{V_i \bar{S}_d} = F_{V_i V_j}^{V_i} (\{\bar{S}_1\}) + F_{V_i V_j}^{V_i} (\{\bar{S}_2\}) + F_{V_i V_j}^{V_i} (\{\bar{S}_1, \bar{S}_2\}) \quad (17)$$

Assuming the subset collections formed by $F_{V_i, V_j}^{V_i, S_d}$ reaches $\bar{S} = \{\bar{S}_1, \bar{S}_2\}$ are $\{\bar{S}_1\}$ and $\{\bar{S}_2\}$.

$$f_{V_i V_j}^{V_i \bar{S}_d} = F_{V_i V_j}^{V_i} (\{\bar{S}_1\}) + F_{V_i V_j}^{V_i} (\{\bar{S}_2\}) \quad (18)$$

In that case that node V_i arrives at the collection of all base stations through the adjacent node collection V_j ; that is, all subset collections are formed by $\bar{S} = \{\bar{S}_1, \bar{S}_2\}$

$$\sum_{V_j \in X, X \subseteq S_i} f_{V_i V_j}^{V_i \bar{S}_d} = F_{V_i V_j}^{V_i} (\{\bar{S}_1\}) + F_{V_i V_j}^{V_i} (\{\bar{S}_2\}) + F_{V_i V_j}^{V_i} (\{\bar{S}_1, \bar{S}_2\}) \quad (19)$$

It is assumed in this thesis that the subset collection of data transmitted by node V_i arrives at the base station is Ω with $\Omega \subseteq P$ and $\Omega = \{P_1, P_2, \dots, P_k\}$. $\phi_{V_i}(\Omega)$ Represents the total rate of data transmitted by node V_i to Ω within a unit of time.

$$\phi_{V_i}(\Omega) = \sum_{P_k \subseteq \Omega, P_k \in P} F_{V_i V_j}^{V_i} (P_k) = \sum_{V_j \in X, X \subseteq S_i} \sum_{P_k \subseteq P} F_{V_i V_j}^{V_i} (P_1 + P_2 + \dots + P_k) \quad (20)$$

The data transmitted by node V_i comes from the sum of data received and the data generated by its induction, and the data received by the node Vi from the adjacent node is

$\sum_{V_i, V_j \in \bar{S}, V_j \subseteq X} f_{V_i V_j}^{V_i, S_d}$ and $\phi_{V_i}(\Omega)$ represents the total rate of data received by a node within a unit of time.

$$\phi_i(\Omega) = \sum_{V_j \in X, X \subseteq S_i} \sum_{P_k \subseteq P} F_{V_i V_j}^{V_i} (P_k) = \sum_{V_j \in X, X \subseteq S_i} \sum_{P_k \subseteq P} F_{V_i V_j}^{V_i} (P_1 + P_2 + \dots + P_k) \quad (21)$$

In this chapter, the value of $\phi_{V_i}(\Omega)$ generally is more than that of $\phi_{V_i}(\Omega)$; given $\Psi_i(\Omega) = \phi_i(\Omega) - \phi_{V_i}(\Omega)$ and in the case of $\Psi_i(\Omega) > 0$. In addition to forwarding data within a unit of time, the node Vi also induces and generates data, and $\Psi_i(\Omega) = 0$ shows that the node V_i only forwards data and $\Psi_i(\Omega) < 0$ shows that there is data loss in node Vi or it conducts redundancy processing and data compression for the data[18]. It can be determined from Equation (7) that when the data rate in the node decreases, the energy consumption of the node will also decrease, and when there is no data passing the node, it is considered that the node only needs to maintain relatively low energy consumption, as shown in Figure 3.

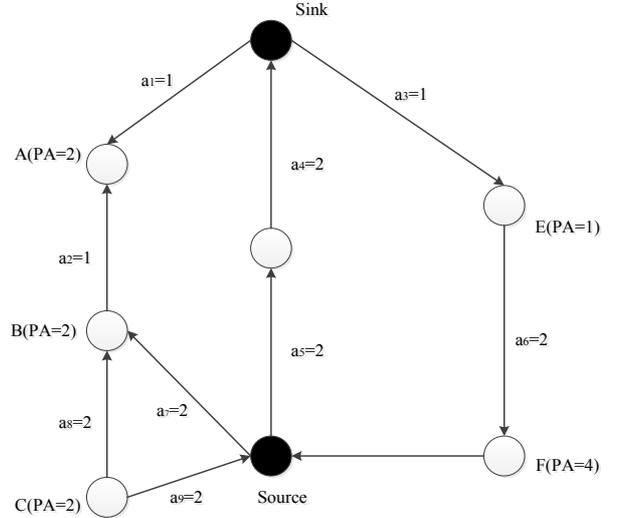


Figure 3. Energy routing protocol

C. Systematic mathematic model

In this chapter, a multi-constraint linear programming model is proposed in order to resolve the maximum lifetime communication model of the wireless sensor network, with network lifetime being its optimal objective. A distributed algorithm is a typical algorithm adopted to solve linear programming model. In order to resolve the maximum lifetime model of a network with adoption of distributive algorithm, substitution variable q is introduced, and set $q=1/T_{sys}(f)$. The solution to the maximum lifetime of a wireless sensor network becomes the solution to minimization of q, and such a model can be transformed into an equivalent model.

$$\text{maximize } q \quad (22)$$

$$\text{subject to } \sum_{V_i, V_j \in \bar{S}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i \quad (23)$$

$$= \sum_{V_i, V_j \in \bar{V}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d}$$

$$e_{ij} \cdot \sum_{V_i, V_j \in \bar{V}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i + \quad (24)$$

$$\rho \cdot \sum_{V_i, V_j \in \bar{V}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + e_{other} \leq q E_i, \forall V_i, V_j \in \bar{V}, \forall \bar{S}_d \in \bar{S}$$

$$\sum_{\bar{S}_d \in \bar{S}} f_{V_i V_j}^{V_i \bar{S}_d} \leq R_{ij}, \forall V_i, V_j \in X, X \subseteq S_i \quad (25)$$

$$f_{V_i V_j}^{V_i \bar{S}_d} \geq 0, \lambda_i = 0 \text{ or } 1, g_i \geq 0, \forall V_i, V_j \in X, X \subseteq S_i \quad (26)$$

The objective function q is technically not a convex function and the dual function is non-differentiable. To solve this problem, the substitution scheme proposed in literature [19] is adopted, where q is substituted with q^2 , and such substitution has no influence on the solution of maximum value of lifetime T. Meanwhile, in this chapter, a differential convex secondary coordination item $\phi(x)$ is introduced based on the characteristics of a wireless sensor network. By setting as $\phi(x)=\Psi(\Omega)$ the KKT condition is satisfied. In order to make $\phi(x)$ become differential or technically a convex function, substitute $\Psi_i(\Omega)$ with $\sigma^2 \Psi_i^2(\Omega)$, where σ is a small positive real number; the linear programming model will be:

$$\text{maximize } q^2 + (\phi_i(\Omega)) \quad (27)$$

$$\text{subject to } \sum_{V_i, V_j \in \bar{S}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i \quad (28)$$

$$= \sum_{V_i, V_j \in \bar{V}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d}$$

$$e_{ij} \cdot \sum_{V_i, V_j \in \bar{V}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + \lambda_i g_i + \quad (29)$$

$$\rho \cdot \sum_{V_i, V_j \in \bar{V}, V_j \subseteq X} f_{V_i V_j}^{V_i \bar{S}_d} + e_{other} \leq q E_i, \forall V_i, V_j \in \bar{V}, \forall \bar{S}_d \in \bar{S}$$

$$\sum_{\bar{S}_d \in \bar{S}} f_{V_i V_j}^{V_i \bar{S}_d} \leq R_{ij}, \forall V_i, V_j \in X, X \subseteq S_i \quad (30)$$

$$\phi_i(\Omega) = \sum_{V_j \in X, X \subseteq S_i} \sum_{R_k \subseteq P} F_{V_i V_j}^{V_i} (P_1 + P_2 + L + P_k) \quad (31)$$

$$\Psi_i(\Omega) = \phi_i(\Omega) - \phi_i(\Omega) \quad (32)$$

$$f_{V_i V_j}^{V_i \bar{S}_d} \geq 0, \lambda_i = 0 \text{ or } 1, g_i \geq 0, \forall V_i, V_j \in X, X \subseteq S_i \quad (33)$$

Equation (19) is the linear programming model described in the optimization theory, which can be resolved with the adoption of the distributed projected gradient algorithm. The approximate upper limit value of the maximum network lifetime can be solved through iterative computation.

IV. SIMULATION AND PERFORMANCE ANALYSIS ON THE ALGORITHM

In order to verify performances of the entire model, simulation has been conducted in NS-2 environment, and the simulation result was applied based on WSTP and MD-MLR to realize random generation in NS-2 environment of 2-6 sink nodes and 30-100 common sensor nodes which are randomly distributed within a planar area measuring 80X80M2. All these nodes are equipped with positioning function and their transmission power is adjustable. R, radius of maximum transmission range, is 15; data generation rate $g_j=1\text{kbit/s}$; data package size is 512Bytes; metadata size is 25Bytes; primary energy transmission unit and data energy consumption factor of sensor node are subject to comparison. $\sigma=50\text{nJ/b}$, $\beta=0.0013\text{pJ/b/m}^4$, and data energy consumption factor of the receiving part $\rho=50\text{nJ/b}$; path power consumption model parameter $m=4$; correction factor of secondary optimization model $\sigma=0.01$.

Figure 4 shows the relationship between total energy consumption of the network and the number of nodes. The number of base station is respectively set as 2, 3, 4, 5 and 6. As the number of network nodes increases, node intensity increases. Meanwhile, the number of source nodes increases proportionally. In fact, when the number of nodes doubles, the data size generated in the network doubles correspondingly. It is obvious that as node intensity increases, the increased energy consumption will be less than doubled. However, the total energy consumption of the network will rise. As the number of base station increases, the total energy consumption rate of the network reduces gradually. With number of nodes being the same, the more the base station there is, the closer is the link formed by source node data to the base station, resulting in less energy consumption. When the number of base station increases, the reduction of average energy consumption of the network caused by the newly accessed base station becomes less. In such a case, as $n=30$, when the number of base station expressed as m, increases from 2 to 4, energy consumption will be reduced about 32%. In other words, when the number of base station doubles, reduction in energy consumption will not exceed 50%. When m is increased to 6, the reduction in energy consumption is such that it is about 48% of that when $m=2$, and 16% further less than when $m=4$. The influence of the number of base stations on the total network energy consumption reduces gradually, while the total network energy consumption shows a declining trend.

Figure 5 shows the comparison between the proposed algorithm MLCM and WSTP and MD-MLR algorithm in terms of maximum network lifetime when there are three base stations in the network. It can be observed from Figure 5 that the algorithm proposed in this chapter is superior to the other two models in terms of network lifetime. According to the MLCM algorithm, data

generated at the source node can be transmitted, in the process of data routing, via the optimal link among many links, or otherwise transmitted via all these links, and the number of such links may increase when data is transmitted from the source node. The MLCM model tries to cause all nodes in the network to participate in the data transmission in order to make the network energy evenly consumed by more nodes, although nodes participating in the transmission may not be located within the task section of the WSTP algorithm. For the WSTP algorithm, the network is divided into sections based on the number of base stations, and each section is under the charge of a certain base station. It is in fact a convergence of a single-base-station network, which is very close to the shortest path model of common with a single base station in regards to energy consumption. In this case, energy consumption is likely to take place at a certain link, making nodes on such link die excessively quickly, which results in a short link lifetime. The MD-MLR algorithm is an event-driven routing model and is the routing algorithm based on minimum network delay and maximum network lifetime. According to this algorithm, route discovery will be re-conducted prior to each routing. The algorithm seeks to improve the efficiency of transmission to the base station and reduce delay. However, the path of data transmission fails to realize the optimal minimum energy consumption path on general network energy consumption, making the optimization unsatisfactory.

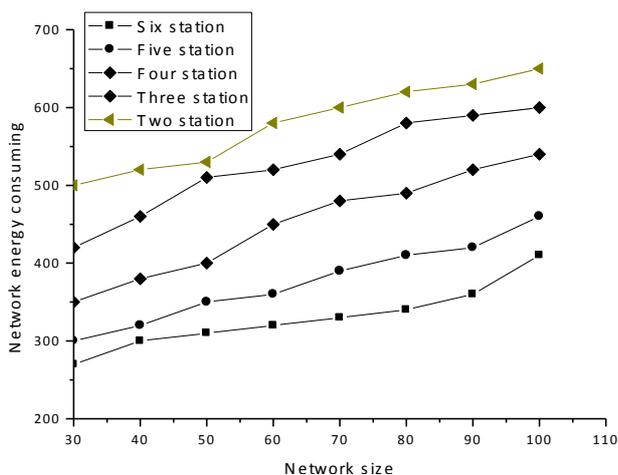


Figure 4. Network energy consumption

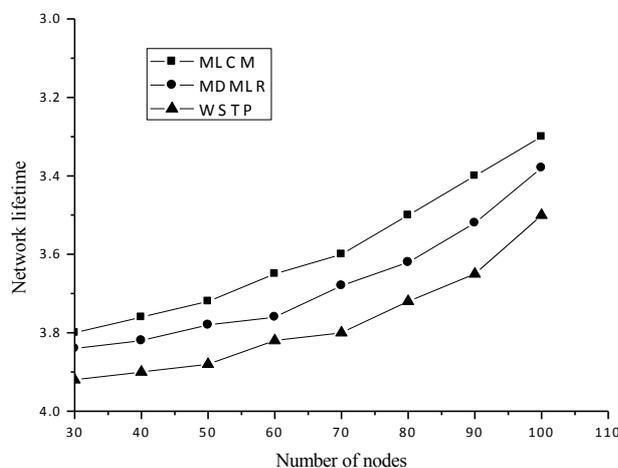


Figure 5. Comparison of network lifetime

V. CONCLUSION

In this thesis, research has been conducted on the maximum lifetime communication model of a multi-source multi-link multi-base station of a wireless sensor network. Multiple base stations are often deployed in wireless sensor networks in order to prolong lifetime, reduce data delay and reinforce network robustness. In terms of wireless sensor network equipped with multiple base stations, data generated at the source node can be transmitted to a number of base stations via multiple links. In the process of data transmission, the primary objective is to realize maximum network lifetime, and the secondary objective is to reduce data size at the node; namely, to make data arrive at the base station via the minimum number of nodes in the network. A linear programming model which takes a maximum network lifetime as the optimal objective is established in order to cope with many constraints such as limited energy and bandwidth of a wireless sensor network. Such a linear programming model is resolved with the adoption of a distributed heuristic algorithm. Finally, performances of the communication model have been verified through a simulation experiment, which shows that the proposed model is effective in improving network lifetime.

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