Keywords: Key Management, Random Key Predistribution, Sensor Networks.

Abstract: The majority of studies on security in resource limited wireless sensor networks (WSN) focus on finding an efficient balance among energy consumption, computational speed and memory usage. Besides these resources, time is a relatively immature aspect that can be considered in system design and performance evaluations. In a recent study (Castelluccia and Spognardi, 2007), the time dimension is used to lower the ratio of compromised links, thus, improving resiliency in key distribution in WSNs. This is achieved by making the old and possibly compromised keys useful only for a limited amount of time. In this way, the effect of compromised keys diminish in time, so the WSN selfheals. In this study we further manipulate the time dimension and propose a deployment model that speeds up the resilience improvement process with a tradeoff between connectivity and resiliency. In our method, self healing speeds up by introducing nodes that belong to future generations in the time scale. In this way, the duration that the adversary can make use of compromised keys become smaller.

1 INTRODUCTION

The significance of wireless sensor networks (WSNs) is that the cheapest possible node model is targeted due to nature of the network; such as being in a hostile environment, unattended, and due to the geographic constraints which prevent reusability of a sensor node. Moreover, the application fields of WSNs, like battlefield surveillance and habitat monitoring need security precautions in order to work as intended (Akyildiz et al., 2002).

For secure communication in WSNs, the symmetric encryption is preferred for the sake of energy consumption and faster processing. For this purpose, the distribution of symmetric keys is obligatory and its difficulty is the main problem of secure communication in WSNs.

Thinking of the very intuitive but inefficient scheme where all possible pairwise keys in the network are kept in the memory of each node, the connectivity is at its utmost. However, the extra memory that is wasted for unused keys, is too expensive for a tiny sensor node.

On the other hand, using a single key in the whole network will be the most desired choice for memory. However, all the links will be compromised if the key is captured by adversary. Therefore, a viable solution would look for an equilibrium in resource consumption in order to deal with the strict constraints. In this sense, Eschenauer and Gligor (Eschenauer and Gligor, 2002) devise a mechanism in which all nodes are given a random amount of keys from the key pool. This scheme results in reasonable levels of connectivity and resiliency.

Besides the advantage of randomness, the time dimension is a reality for any network and should be considered in the design. Castelluccia and Spognardi proposed RoK (A Robust Key Pre-distribution Protocol for Multi-Phase WSNs) (Castelluccia and Spognardi, 2007), that takes time dimension into account. In RoK (Castelluccia and Spognardi, 2007), the network lifetime is divided into phases. At the end of each phase, all the keys of nodes are updated. Thus, new links are established with updated keys that are out of the reach of adversary before compromising nodes with new keys.

In this study, we improve the resiliency of RoK by further exploiting the time dimension. Our contribution is to use keys that are assigned to future uses, earlier than their times. As a result, we end up with improved resiliency measures. As explained in Section III, we propose two models called Constant Offset Future Random Generations (COFRG) and Grow-


ing Offset Future Random Generations (GOFRG). At each deployment phase, both of them choose a time interval in future. Some of the keys from this time interval are chosen randomly and used in the current time. In COFRG, the time interval is a fixed offset from current time. However, in GOFRG this interval has growing offsets with respect to present. In this way, at each deployment of GOFRG, a high fraction of deployed keys become new to adversary. This is valid for COFRG too, but the fraction of new keys at each deployment of COFRG becomes lower after the initial stages of the network. Therefore, the contribution of GOFRG to resiliency is better as compared to COFRG. On the other hand, connectivity decreases in both models due to higher number of nodes that belong to future generations in comparison to RoK.

The rest of the paper includes the background information on Multi-Phase Keying in Section II. In Section III the proposed schemes are presented. The advantages versus the drawbacks of the schemes are discussed in Section IV. Related works are described in Section V. Finally Section VI summarizes the conclusions.

2 PREVIOUS WORK IN MULTI-PHASE KEYING

In Multi-Phase Keying models, the network life is divided into phases of equal time intervals. All the keys in the key pool and in the key rings of nodes are updated with each phase, such that the adversary fails to derive the keys of future phases from previously captured keys. At the same time, deriving keys of previous phases from current phase keys is prevented. However for the sake of connectivity among nodes that are active in neighboring phases, this prevention mechanism is activated gradually, as explained below. All the schemes that are discussed in this paper, namely RoK, COFRG and GOFRG are different variations of the multi-phase approach.

Each phase is called a generation which consists of 10 rounds, where one round is the smallest unit of time. The reason for this time segmentation is that the attack scenarios are based on rounds.

In RoK (Castelluccia and Spognardi, 2007) all the keys are identified with the generation in which they are used. So that, by the end of each generation all the valid keys are updated. However, in order to compromise the maximum number of links, an attacker may prefer to update the key ring of each captured node forever. To prevent this, a security mechanism should be able to guarantee that the key ring of any node is bound to a given amount of time. After exceeding this time, a node should no more establish a secure communication between new deployed nodes. In RoK (Castelluccia and Spognardi, 2007) this time duration is set to 10 generations, which is almost the maximum battery life of a node. This binding is provided by the backward and forward hash chains.

As a result of this binding, the keys obtained from captured nodes get old by time and new established links remain safe. This decreases the ratio of compromised links with every generation, if adversary stops capturing new nodes. The decrease in this ratio, i.e. the improvement in resiliency, is also called the self-healing of the network.

2.1 Node Configuration Phase

At the beginning of each generation, a set of sensor nodes are deployed with forward and backward key rings. These key rings are hashed at the end of each generation, so that the new key rings are identified with the new generation. This way nodes maintain their lives among generations. In this scheme, forward and backward hash chains, constitute the update mechanism mentioned above, satisfying its security requirements thanks to the irreversibility of hash functions.

Each element of the forward and backward hash chains will be referred as a Forward Key or Backward Key. The key rings are sets containing a number of chosen Forward Keys and Backward Keys from the pools, called Forward Key Pool and Backward Key Pool. The Forward Key Pool, at Generation 0, i.e. the first deployment of the network, is defined as follows. Please refer to Table 1 for the definitions of symbols:

\[
FKP^0 = f k_1^0, f k_2^0, \ldots, f k_p^{0/2},
\]

where each \( f k_i^0 \) is randomly generated.

At Generation \( j + 1 \), the forward keys are refreshed as follows:

\[
f k_i^{j+1} = H(f k_i^j),
\]

\[
FKP^{j+1} = f k_1^{j+1}, f k_2^{j+1}, \ldots, f k_p^{j+1}
\]

The Backward Key Pool, is first generated for Generation \( n \), i.e. the last generation of the network. The backward keys at Generation \( n \), are initialized with random values:

\[
BKP^n = b k_1^n, b k_2^n, \ldots, b k_p^{n/2}.
\]

At Generation \( j \), the backward keys are refreshed as follows:
Table 1: Symbols Used in Multi-Phase Keying.

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>P</td>
<td>Key Pool Size</td>
</tr>
<tr>
<td>m</td>
<td>Key Ring Size</td>
</tr>
<tr>
<td>FKP</td>
<td>Forward Key Pool</td>
</tr>
<tr>
<td>BKP</td>
<td>Backward Key Pool</td>
</tr>
<tr>
<td>fk</td>
<td>Forward Key</td>
</tr>
<tr>
<td>bk</td>
<td>Backward Key</td>
</tr>
<tr>
<td>gx</td>
<td>The generation of node X</td>
</tr>
<tr>
<td>X_i</td>
<td>Item X with Generation j and index i</td>
</tr>
<tr>
<td>LT</td>
<td>Life Time of the key ring of a node.</td>
</tr>
<tr>
<td>H', H''</td>
<td>Two different hash functions.</td>
</tr>
<tr>
<td>jk_l-bk_l</td>
<td>A forward-backward key pair.</td>
</tr>
</tbody>
</table>

\[ bk_l^j = H'(bk_l^{j+1}) \]
\[ BKP_l^j = bk_l^j, bk_l^{j+1}, ..., bk_l^{j+p/2} \]

Therefore, at Generation \( j+1 \) the backward key pool is defined as:
\[ BKP_l^{j+1} = bk_l^{j+1}, bk_l^{j+2}, ..., bk_l^{j+p/2} \]

Every node is configured with forward and backward keys in the following way: For a node with ID_A and generation \( g_A \), the \( i^{th} \) key of the Forward Key Ring is the key from the Forward Key Pool of index \( H'(ID_A|i|g_A) \). This is done for all \( m/2 \) keys in the Forward Key Ring. For the Backward Key Ring the same operation is performed using the indices of the Backward Key Pool.

### 2.2 Key Establishment Phase

After deployment, a node A broadcasts ID_A and its generation, \( g_A \). A receiver node B, at first, decides if their generations are close enough or not. This is done by testing if \( |g_A - g_B| < LT \). In addition to this, if \( g_A < g_B \) and the above holds, then, they can share keys starting from Generation \( g_B \) up to Generation \( g_A + LT - 1 \).

Secondly, Node B calculates \( H'(ID_A|i|g_A) \) and compares them with its indices, \( H'(ID_B|j|g_B) \) for all \( i, j \in 1,2,m/2 \). If there are collisions such that

\[ H'(ID_A|i|g_A) = H'(ID_B|j|g_B) \]

where \( x,y \in 1,2,..m/2 \), then, it is known that they both have the forward key \( f_{H^l(ID_A|i|g_A)} \) and the backward key \( b_{H^j(ID_B|j|g_B)} \) in their memory. This way, all colluding local indices \( a,b,z \in 1,2,..m/2 \) are found and the following becomes their pairwise symmetric key:

\[ K = H'(f_{H^l(ID_A|i|g_A)}[b_{H^j(ID_B|j|g_B)}[H^a(ID_A|i|g_A)]|...][b_{H^j(ID_B|j|g_B)}[H^a(ID_A|i|g_A)]|...]) \]

Any attacker needs all these forward and backward keys to compromise this pairwise key. These keys can not be reached using a particular forward–backward key pair. A forward key is reachable only through a suitable past forward key and a backward key is reachable only through a suitable future backward key; these suitable keys need to have the same indices with the keys in 9. Therefore, an adversary would construct a table that is filled with the hash chains of the captured keys. This way future forward keys and past backward keys can be calculated using the hash function as in 2 and 5.

### 3 PROPOSED SCHEMES

The hashing mechanism in RoK (Castelluccia and Spognardi, 2007) and its usage of time dimension through generations provide the self healing ability of the network. In our study, we modify the node deployment model of RoK by using nodes of future generations. Therefore the network acts as if it has the state of a few generations later, which results in a faster self healing process.

In this study, we propose to use nodes that belong to a random future generation, at each deployment. This method will be referred as Future Random Generations. Two different models on how to choose from future generations are proposed as explained below. In the classical RoK approach, the attacker is able to compromise keys of established links provided that the captured nodes and the link that is to be captured have overlapping generations. In the proposed models, we enable a faster self healing and improve resiliency by reducing the probability of overlapping generations via future random generations.

At the end of each generation, some of the nodes including the newly deployed ones have key rings that belong to a few generations ahead. In this way, each node in the network happen to live in a different generation than most of its neighbors. Therefore, this can be referred as a generation mixture or traveling in time.

The early deployment of Future Random Generations would cause a decrease in connectivity. Actually our method creates a tradeoff between resiliency and connectivity, which is analyzed in Section IV.

### 3.1 Deployment Models

In RoK (Castelluccia and Spognardi, 2007) at each generation, the new nodes that are deployed over the
field are chosen such that they belong to current generation. However in our schemes, the generations of the new nodes are chosen randomly. The range of generations from which the generation of each new node is randomly selected is defined as deployment window. The position of the deployment window on the time scale shifts towards future at each generation. The rules of shifting the deployment window constitute our deployment models. We propose two such models, namely COFRG (Constant Offset Future Random Generations) and GOFRG (Growing Offset Future Random Generations), that are detailed in the following subsections. In both models, the size of the deployment window is fixed to 10 generations.

In our models, each new node is assigned a uniformly random generation picked out of the current deployment window.

### 3.1.1 Constant Offset Future Random Generations (COFRG)

In COFRG, the deployment window has a constant offset to current generation. The deployment window shifts one by one at each generation. In this way, the offset between the deployment window and the current generation remains unchanged.

In COFRG, the network is initialized without considering the deployment window rules and all the nodes are deployed as Generation 0 nodes. However, all nodes to be deployed after Generation 0 have generations randomly selected out of deployment window.

The discrete uniform random variable that determines the generation of a specific node, $G_{COFRG}$, is defined as follows.

$$G_{COFRG} = \begin{cases} 0 & \text{if } T = 0 \\ T + D + X & \text{if } T > 0 \end{cases}$$

where $X$ is a random integer uniformly distributed in \{0, 1, ..., 9\}. $T$ is the index of current generation and $D$ is the offset to current generation.

At Generation $T$, the deployment window covers the range $T + D$ to $T + D + 9$. The generation of each node to be deployed is a uniform random variable, $G_{COFRG}$, picked out of this deployment window. In the next generation, $T + 1$, the deployment window is shifted one step forward having the range $T + 1 + D$ to $T + 1 + D + 9$. The generation of all nodes to be deployed at $T + 1$ is selected randomly from this deployment window. This goes on for all consecutive generations.

Fig. 1 exemplifies both deployment window and the existing generations on the field in COFRG with $D = 5$. Each cell with dotted background is a deployment window and the symbols in these cells represent the range of generations in that deployment window. The horizontal axis shows the current generation. For an example, the deployment range of Generation 4 is between the generations 9 and 18 (inclusive). A node that is to be deployed in the current generation is assigned a future random generation out of the deployment window that corresponds to this current generation.

The vertical axis is a reference to observe the existing generations on the field. In addition to dotted background that corresponds to deployment window of current generation, the generations with red grid texture show the ones that have been deployed prior to current generation. For example, the deployed generations at the time of Generation 3 are Generation 0 and the generations between 6 and 17.

The generations between 1 and 5 are never deployed in any generation. This is due to the constant offset feature of COFRG.

### 3.1.2 Growing Offset Future Random Generations (GOFRG)

In GOFRG, the deployment window shifts towards future with some jumps. Each node is assigned a generation which is determined by a discrete uniform random variable, $G_{GOFRG}$, as follows.

$$G_{GOFRG} = \begin{cases} 0 & \text{if } T = 0 \\ (T - 1) \ast JUMP + T + X & \text{if } T > 0 \end{cases}$$

where $X$ is a random integer uniformly distributed in \{0, 1, ..., 9\}.
\{0,1,\ldots,9\}$. $T$ is the index of current generation and $JUMP$ is the length of additional offset.

![Diagram](image_url)

**Figure 2**: Deployed Generations vs. Generations of Deployment in GOFRG.

Besides the natural increase in the time scale (one by one), the deployment window in GOFRG makes additional shifts with the length of $JUMP$ at each generation. Hence, a deployment window of GOFRG increases its offset to current generation with constant speed. The $JUMP$ parameter is constant for a given GOFRG model.

Fig. 2 illustrates the deployment windows and existing generations up to the fifth generation of the network in GOFRG with $JUMP=2$. The deployment range at Generation 3 is between 7 and 16. In this case the deployment has an offset of 4 to the current time. The following generation (Generation 4) has the deployment window with range 10 to 19, which has an offset of 6 to the current generation. In this way, the difference between the deployment window and current generation increases as generations go by.

For each deployment in both COFRG and GOFRG, the links established using generations that are deployed for the first time cannot be compromised using the nodes captured in previous generations. The number of these safe generations is $(JUMP + 1)/10$ of the length of the deployment window. This ratio is 1/10 in COFRG, considering that $JUMP = 0$. Therefore, as compared to COFRG, higher fraction of generations are out of the reach of adversary in GOFRG. This difference leads to higher resiliency values for GOFRG as will be discussed in 3.3.

In COFRG scheme, the offset is kept constant in order not to be too far from current generation. Consequently, connectivity is kept within reasonable levels. However, this balance between offset and connectivity is not taken into consideration in GOFRG for the sake of better resiliency.

### 3.2 Key Establishment Phase

The key establishment phases of both models COFRG and GOFRG are identical with RoK, however the results are different as explained in 3.3. The generation overlaps in COFRG and GOFRG are fewer compared to RoK. The reason is that, the deployment generations are mostly chosen from future time domains, so the generation overlap probabilities between the key rings of nodes are reduced. Therefore, less node pairs are able to establish shared keys, however, the resulting key becomes more resilient in GOFRG than RoK, as explained in 3.3.

As in most of the WSN applications, whenever two neighboring nodes are not able to establish a pairwise key using the key rings in their memories, they apply path key establishment procedure in order to communicate in a secure way. The path key establishment phase has the following steps:

1. One of the nodes broadcasts a message that contains the IDs of the two nodes in question, looking for an anchor node that shares a key with both of the nodes.
2. This broadcast is flooded across the network until it reaches an anchor. This step will increase the communication overhead of the nodes involved. Therefore the broadcast is allowed to make at most, say, 3 hops.
3. The anchor node generates a random pairwise key for the two nodes and sends it to both parties using the secure channels established earlier.

The path key establishment is supposed to keep the connectivity in COFRG and GOFRG in desired levels, with a cost of energy consumption due to communication overhead. However, the positive effects of path key establishment on connectivity are not shown in the figures below in order to observe the connectivity prior to path key establishments.

### 3.3 Performance Evaluation

RoK scheme (Castelluccia and Spognardi, 2007) explains in detail how Multi−Phase Keying mechanism improves resilience over time. This behavior is called the self healing ability of the network, which addresses the decrease of adversary ability to compromise new links with a given number of captured nodes. As a result the fraction of compromised active links used in the network decreases.
Since our goal is to speed up the self healing process and observe the resulting resiliency and connectivity metrics, by employing the proposed Future Random Generations approach, the detailed comparison between previous schemes which was done in (Castelluccia and Spognardi, 2007), is not repeated in this paper.

3.3.1 Simulation Details and Performance Metrics

The simulations were implemented in C# .Net 2005 on Windows XP SP2. Each simulation was run 20 times for the sake of accuracy. COFRG and GOFRG schemes were tested together with RoK. For simplicity 20*20 area is used to deploy 400 nodes. With vertical and horizontal neighboring, each node has exactly 4 neighbors.\(^1\)

At the end of each generation, the nodes that run out of battery are replaced with new nodes, which are configured according to the rules of the related scheme. This replacement obviously is not feasible in real life but to cope and compare the results with RoK scheme, a similar deployment is adopted.

For all scenarios, the sizes of both forward and backward pools are 100,000 and the sizes of forward and backward rings of a node are both 100.

The lifetimes of nodes are decided according to Gaussian distribution with mean 5 and standard deviation 10/6.

The simulations were run with two attacker models: In the first group, the attacker, called the constant attacker, captures 5 nodes per round. In the second, the attacker, called the temporary attacker, captures nodes only until the end of Generation 9, again with a rate of 5 nodes per round.

The figures below show two kinds of measurements, the compromise ratio and the local connectivity for all the models RoK, COFRG and GOFRG. For the calculation of the compromise ratio, all links that are compromised by adversary are counted except the links that belong to captured nodes. This count is divided by the total of all links that belong to non captured nodes. In addition, this ratio was calculated separately for active and total compromised links in order to differentiate between the compromise of active and dead links. Noting that the compromise ratio is the inverse of resiliency metric and the drop in compromise ratio implies the increase in resiliency and visa versa. Here a dead link refers to a link which has at least one of its end nodes has gone out of battery. An active link is visa versa, i.e. both of its ends have enough battery to communicate.

For local connectivity, the key establishment requests between neighboring nodes are counted. In these key establishment attempts, the number of successful ones that end up with valid key establishments were divided into the total of all the attempts. The result show the amount of success of the related scheme in terms of local connectivity. Despite that low connectivity is supported by path key establishments, this support is not reflected the graphs below in order to observe the connectivity performances of all schemes.

3.3.2 Simulation Results

Fig. 3 shows the number of all compromised links over all the links established since the beginning of the network versus generations, with the constant attacker model. Here, at the early stages of the network the adversary is able to benefit from the captured nodes and increase the compromise ratio immediately, which is due the majority of Generation 0 nodes in the area. After this early dramatic increase until around Generation 5, all the schemes change their behavior. The reason is that by Generation 5 the majority of the nodes scattered in Generation 0 are out of battery and replaced by new nodes. In this way, all the schemes start to follow a steady behavior. At the beginning of the network, all schemes record above 0.4 compromise ratios. After that, a significant drop in compromise ratio, implying the improve in resiliency, for COFRG and GOFRG schemes can be seen; where GOFRG with Jump 3 reaches around 0.2 compromise ratio. Finally, the resiliency of all schemes begin to drop slowly until the end of the network due to the compromise rate of 50 nodes per generation, despite this, multi-phase approach prevents the adversary to go beyond 91% of compromise ratio at worst case.

\(^1\)These parameters are kept the same as (Castelluccia and Spognardi, 2007).
In RoK (Castelluccia and Spognardi, 2007), there is no generation mixture, so at each new deployment only keys belonging to a single generation are introduced to the network. Therefore, they are certainly unknown to adversary at the time of deployment. On the other hand, for each deployment of COFRG, after the network reaches a steady state by Generation 5, the generations of the nodes that are deployed contain already deployed generations with ratio 9/10. In other words, only 1/10 of the deployed nodes are from generations that do not exist in the area yet. This causes high compromise ratios in the latter stages. However it has around 0.3 compromise ratio at Generation 5 while RoK records 0.45 at that time. This advantage for the resiliency in COFRG is due to having Offset 5 from current time and the majority of nodes on the area being of Generation 0 (see Fig. 1).

The only difference of GOFRG compared to COFRG is the JUMP parameter which is 0 in COFRG and has larger values for GOFRG. The performance of both schemes is similar until Generation 5, where a significant drop in compromise ratio is achieved. In order to maintain this performance, at each deployment, extra jumps towards future is made by GOFRG. This results in a better resiliency performance than both of COFRG and RoK throughout the network life.

Meanwhile, using future generations in COFRG and GOFRG pays off with lower connectivity performance (Fig. 4). This is due to the nodes from future generations that have lower probabilities of having colliding generations with their neighbors compared to RoK. Fig. 1 and Fig. 2 show the generation diversity at a given time. As it is seen in Fig. 1, with COFRG at Generation 5 the generations between 0-19 exist in the network. Therefore it is more difficult for COFRG nodes to have colliding generations between their neighbors, according to RoK which has nearly 10 generations at a given time, considering the battery lifetimes of nodes. The same applies for GOFRG, where the diversity of generations causes loss in connectivity too. In Fig. 2, at Generation 5 there are 22 generations ranging from Generation 0 up to Generation 22. However, the low connectivity is tolerated with path key establishments which increase the communication overhead. Despite this communication overhead, the tradeoff between connectivity and resiliency is desirable since resiliency has no alternative.

Fig. 5 shows the compromise ratio of active links, which are certainly more valuable than the dead links for most of the applications of WSNs. The compromise ratio of all schemes oscillate with certain equilibrium and do not exceed certain limits, despite the capture rate of 5 nodes per round. In Fig. 5, the low compromise ratio of GOFRG throughout the whole network life, compared to RoK and COFRG shows that, its high resiliency values is also valid for active links. During the steady state of the network, COFRG is around 0.7 of compromise ratio and RoK oscillates between 0.55 and 0.4. However, GOFRG perform better with oscillations between 0.21 and 0.45.

The temporary attacker in Fig. 6, does not compromise any nodes after Generation 9. In addition to that, the adversary can compromise new links until Generation 19 in COFRG, which is the extremum case. In GOFRG, adversary can not compromise links after Generation 12. Later on, the compromise ratio
that does not reach 0 is due to the dead links at the hand of adversary that are taken into account in Fig. 6.

Meanwhile, Fig. 7 is decisive in terms observing the self healing abilities of all the schemes. Fig. 7 shows for all schemes that the number of compromised active links becomes 0, which implies that the self healing of all schemes manage to heal the network completely. However, GOFRG Jump 3 and GOFRG Jump 2 reach 0 compromise ratio by Generation 14. While, GOFRG Jump 1 and RoK achieve it by Generation 15. Finally, COFRG achieve the same level by Generation 18. Noting that these statistics are not the only difference between the schemes, since until the complete self healing achievement of the schemes, high resiliency values of GOFRG against RoK and COFRG is notable.

4 DISCUSSIONS

The faster self healing of GOFRG is observed, however this model gives us a wide range of new issues to consider, like the tradeoff between connectivity and resiliency, dynamic lifetime of key rings and uncaptured nodes that turn to be useless. In this section we discuss these issues.

Deploying some of the nodes earlier than their own generations obviously will make it harder for the attacker to compromise new links, however it will also make it harder for new nodes to establish keys with older ones. In this case, path key establishments, certainly with some energy cost, bring connectivity to desired levels. Keeping in mind that resilience is rather harder to tolerate, this scenario would be fairly desirable for attack sensitive applications. In these applications, low local connectivity can be balanced by path key establishments in order to achieve reasonable connectivity levels.

In RoK (Castelluccia and Spognardi, 2007) all nodes have a fixed key ring lifetime, $LT$. Since a random mix of generations are deployed in COFRG and GOFRG, those key rings may expire earlier than expected. Since there may not be any colliding generations between the node in question and the neighboring nodes deployed a few generations later. This will cause a waste in sensor nodes that become useless even though they have enough battery to operate. We plan to investigate this issue in a future study.

5 RELATED WORK

In 2002, Eschenauer and Gligor (Eschenauer and Gligor, 2002) proposed a random key predistribution model, for pairwise key sharing of sensor nodes. This study inspired many other researchers to develop random key predistribution mechanisms such as (Du et al., 2003), (Camtepe and Yener, 2004), (Mehta et al., 2005), (Yu and Guan, 2005) and (Anjum, 2006).

On the other hand, a recent deterministic method by Dong and Liu (Dong and Liu, 2007), uses a number of assisting nodes, that are used only for key establishment and correspond to a fraction of 0.8% over all the nodes. The study of Lu et. al. (Lu et al., 2006) is considering the routing mechanism. This work devises a heterogeneous network structure where some of the nodes have extra capabilities in terms of storage, transmission power, etc. Chan and Perrig (Chan and Perrig, 2005) use intermediate nodes for a scalable key establishment scheme, where communication and memory overheads grow sublinearly with the growth of network size.

6 CONCLUSIONS

Despite the limited resources in wireless sensor networks, significant amount of energy and memory are spent for security needs. However, the time dimension is an immature aspect which was not considered to improve performance until recently.

In recently proposed RoK Scheme (Castelluccia and Spognardi, 2007), the key rings of nodes are updated such that older versions of the same key do not reveal the new version benefiting from the irreversibility of hash chain mechanisms. This scheme results in a self healing property of the network that improves resiliency in time.

In this study, we propose to speed up the self-healing process of RoK, which gives better results in terms of resiliency. Two new models, namely COFRG(Constant Offset Future Random Generations) and GOFRG(Growing Offset Future Random
Generations) are described and their performances are analyzed and compared to RoK scheme. Both proposed models make use of generations that are assigned for future uses. COFRG keeps the offset to the current generation unchanged. Meanwhile GOFRG makes jumps towards future in order to increase the offset to current generation. \textit{JUMP} parameter defines the amount of increase in the offset at each generation in GOFRG. In our simulations, the compromise ratio of GOFRG with \textit{JUMP}=3 approaches 0.2 where RoK scheme records more than 0.5 of compromise ratio. That means, GOFRG shows better resiliency as compared to RoK. The local connectivity value for GOFRG with \textit{JUMP}=3 is around 0.5, whereas this metric for RoK is around 0.89. However, local connectivity increases in GOFRG for smaller \textit{JUMP} values with a cost of reduced resiliency. These analyses indicate a tradeoff between connectivity and resiliency in our schemes. This tradeoff is the main difference between the proposed GOFRG scheme and RoK.

The COFRG model, which is actually a special case of GOFRG with zero \textit{JUMP}, is a baseline for resiliency in terms of the \textit{JUMP} parameter. The advantage of GOFRG is that its deployment window shifts more than one generation each time, whereas the deployment window in COFRG shifts one by one. This small difference makes a big effect throughout the network life and resiliency significantly drops in GOFRG. In other words, GOFRG takes the advantage of time dimension in a better way than COFRG.

The advantage of GOFRG in terms of resiliency pays off with low connectivity values. This tradeoff between resiliency and connectivity can be justified considering that connectivity can be tolerated with path key establishments, where low resiliency cannot be cured. We plan to further investigate the time factor on other key distribution methods proposed in the literature.

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