OPTIMIZING CRYPTOGRAPHIC THRESHOLD SCHEMES FOR THE USE IN WIRELESS SENSOR NETWORKS

Position Paper

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Keywords: Sensor networks, Threshold cryptography, Efficient implementation, Multiparty multiplication protocol of Gennaro, Rabin and Rabin.

Abstract: A huge number of small, computationally restricted sensor nodes can be connected wirelessly to form a sensor network. Such networks can be used to monitor large areas and communicate a multitude of measurements (like temperature, humidity, radiation, and so on) to a remote base station. Since this communication happens over the air interface, the transmitted messages are susceptible to forgery, manipulation and eavesdropping. Conventional cryptographic countermeasures against these kind of attacks cannot be readily applied in the context of sensor networks, due to the limited resources of the individual nodes. Since single nodes can be very easily captured and examined, symmetric schemes with the secret key present in every (or at least a subset of) node(s) pose quite a risk in this setting. In this work, we examine the applicability of threshold cryptographic techniques, especially the Gennaro-Rabin-Rabin multiparty multiplication protocol, for sensor networks by employing several optimizations to the different steps of this algorithm, building on previous results we obtained. We are able to improve the running time up to a factor of 6 compared to an unoptimized version for a bitlength of 1,024 Bit and 33 players.

1 INTRODUCTION

Wireless Sensor Networks (WSNs), where a (potentially huge) number of small, resource-constrained sensor nodes is deployed in a large area to measure a wide variety of parameters and communicate them wirelessly in a hop-to-hop manner to a base station for evaluation are still an emerging field of technology. They can be used to efficiently monitor things like water quality, temperature distribution or radioactive particles in areas where approaches using wired devices are too costly or even impossible.

The current challenge when dealing with wireless sensor networks is the difficulty to achieve confidential, authenticated communication over the air interface. Common techniques against eavesdropping, message forgery and manipulation that can easily be deployed on stationary PCs usually do not work in WSNs, due to the huge constraints in terms of available memory, computing power and energy of the individual nodes. The usual way to secure WSNs today is to use symmetric cryptographic techniques, which in general can be calculated much more efficiently than their asymmetric counterparts. The problem with this approach is the storage and distribution of the keys: two sensor nodes can only communicate when they share a common symmetric key. But due to the special structure of WSNs, the loss or malicious removal of single nodes goes largely undetected, so that an attacker can easily try to extract the secret key from a captured node. To avoid the whole network becoming compromised by such an attack, usually only a
certain number of nodes share the same key, which raises the new problem of key distribution. A general overview of different key management techniques usable in WSNs is given in (Merwe et al., 2007).

Given these special challenges, the use of threshold cryptography becomes attractive: instead of storing the secret key on a single node, a number \( t+1 \) of uncompromised nodes must cooperate to generate a valid secret. Capturing a single node is now useless for an attacker, he has to gain access to at least \( t+1 \) nodes to extract the individual shares of the secret and combine them. There is a multitude of threshold cryptography schemes proposed in the literature, their main problem usually being the computational complexity.

In this work we extend our previous work on the subject by further trying to optimize the Gennaro, Rabin and Rabin (GRR) (Gennaro et al., 1998) protocol, thereby improving the applicability of this protocol in the context of sensor nodes.

The remainder of this position paper is structured as follows. Section 2 gives a general introduction to the protocol of Geimaro; Rabin and Rabin and details of our optimizations. Section 3 then presents our current experimental results, while finally Section 4 gives an outlook on our next steps planned.

2 THE PROTOCOL OF GENNARO, RABIN AND RABIN

Classical theoretical results (Ben-Or et al., 1988; Chaum et al., 1988; Goldreich et al., 1987; Yao, 1986) show that any multiparty computation can be performed securely if the number of corrupted participants does not exceed certain bounds. For a survey of these results the reader is referred to the article of Cramer and Damgård (Cramer and Damgård, 2005).

Unfortunately, without further optimizations these results are not easily applicable in real world applications. One of the most prominent examples for the efforts to accelerate these approaches is the paper of Gennaro, Rabin and Rabin (Gennaro et al., 1998). Among other results, it presents a more efficient variant of the Ben-Or, Goldwasser and Wigderson (Ben-Or et al., 1988) multiplication protocol. It gives a protocol for the fast multiparty multiplication of two polynomially shared values over \( \mathbb{Z}_q \) with a public prime number \( q \).

Polynomial sharing refers to the threshold scheme originally proposed by Shamir (Shamir, 1979), which assumes that \( n \) players share a secret \( \alpha \) in a way that each player \( P_i \) (\( 1 \leq i \leq n \)) owns the function value \( f_{\alpha}(i) \) of a polynomial \( f_{\alpha} \) with degree at most \( t \) and \( \alpha = f_{\alpha}(0) \). Then any subset of \( t+1 \) participants can retrieve the secret \( \alpha \) (for example by Lagrange’s interpolation formula). At the beginning of the multiplication protocol each player \( P_i \) holds as input the function values \( f_{\alpha}(i) \) and \( f_{\beta}(i) \) of two polynomials \( f_{\alpha} \) and \( f_{\beta} \) with maximum degree \( t \) and \( \alpha = f_{\alpha}(0), \beta = f_{\beta}(0) \). At the end of the protocol each player owns the function value \( H(i) \) of a polynomial \( H \) with maximum degree \( t \) as his share of the product \( \alpha \beta \). Multiplication protocols of this type are important cryptographic primitives. In particular, they play a decisive role in comparing shared numbers (Damgård et al., 2006) and in the shared generation of an RSA modulus by a number of participants such that none of them knows the factorization (Algesheimer et al., 2002; Catalano, 2005).

The multiplication protocol of Gennaro, Rabin and Rabin (Gennaro et al., 1998) consists of two steps and requires one round of communication and \( O(n^2 k \log n + nk^2) \) bit-operations per player, where \( k \) is the bit size of the prime \( q \) and \( n \) is the number of players.

In step 1. player \( P_i \) (\( 1 \leq i \leq 2t+1 \)) computes \( f_{\alpha}(i)f_{\beta}(i) \) and shares this value by choosing a random polynomial \( h_i(x) \) of maximum degree \( t \), such that \( h_i(0) = f_{\alpha}(i)f_{\beta}(i) \). He then gives player \( P_{j} \) (\( 1 \leq j \leq n \)) the value \( h_i(j) \).

In (Lory, 2007) a modification of this step is given, which reduces its complexity from \( O(n^2 k \log n) \) to \( O(n^2 k) \) (and thus the complexity of the entire protocol to \( O(n^2 k + nk^2) \)) by utilization of Newton’s scheme of divided differences.

However, in many practical situations (e.g. the above mentioned shared generation of an RSA modulus) \( k \) (typically \( k = 1024 \)) will exceed \( n \) and the \( O(nk^2) \)-term will still dominate. For these cases, in (Lory, 2009) a protocol is given, which modifies step 2 to require only \( O(n^2 k) \) bit-operations per player. All of the above mentioned optimizations were also implemented and subsumed in (Koschuch et al., 2010).

In this work, we perform an additional investigation of step 2: in this step, each player \( P_{j} \) (\( 1 \leq j \leq n \)) determines his share \( H(j) \) of \( \alpha \beta \) locally computing the linear combination

\[
H(j) = \sum_{i=1}^{2t+1} \lambda_i h_i(j),
\]

(1)

where the values \( h_i(j) \) have been communicated to him by players \( P_i \) (\( 1 \leq i \leq 2t+1 \)) during step 1. Here, the \( \lambda_i \) are the coefficients of Lagrange’s interpolation formula of degree \( 2t \), which interpolate the support abscissas \( i = 1, 2, \ldots, 2t+1 \) to 0. In general, for a polynomial of degree \( d \) – 1 these known non-zero-
constants are given by
\[ \lambda_i^{(d)} = \prod_{k \neq i} \frac{1}{k - i} \mod q. \] (2)

Expanding this equation becomes:
\begin{align*}
\lambda_i^{(d+1)} &= \frac{1}{(-1)(-2)(-3)(-4)\ldots(-d+1)} \\
&= (-1)^{-d} [d(d+1)(d+2)(d+3)\ldots(d+1-i)].
\end{align*}

Consequently
\begin{align*}
|\lambda_i^{(d+1)}| &= \frac{|d(d+1)(d+2)(d+3)\ldots(d+1-i)|}{2^i} \\
|\lambda_i^{(d)}| &= \frac{|d(d+1)\ldots (d+1-i)|}{2^i} \\
|\lambda_{i-1}^{(d)}| &= \frac{|d(d+1)\ldots (d+1-i)|}{2^i}.
\end{align*}

and
\[ |\lambda_i^{(d+1)}| + |\lambda_i^{(d)}| + |\lambda_{i-1}^{(d)}| = \frac{(d(d+1)\ldots (d+1-i)(d+1-i+1)}{2^i}. \]

From this it follows that for equidistant support abscissas \(i = 1, 2, \ldots, d\) (as they are used in the GRR protocol) the unreduced coefficients \(\lambda_i^{(d)}\) of Lagrange’s interpolation formula of degree \(d = 1\) obey the recursion
\[ |\lambda_i^{(d+1)}| = |\lambda_i^{(d)}| + |\lambda_{i-1}^{(d)}|. \] (3)

This and trivial initial values demonstrate that the \(\lambda_i^{(d)}\) are always integers.

This fact has the consequence that the reduced coefficients as given by Equation (2) can be calculated very easily, because no computation of a modular inverse is necessary. In order to keep the absolute values of the coefficients low, the reduction should not be done into \(\mathbb{Z}_q\) but rather into \(\mathbb{Z}_q := \{x \in \mathbb{Z} | 0 \leq x < q\}\). Rather, the coefficients should be from \(\mathbb{Z}_q := \{x \in \mathbb{Z} | -q/2 < x \leq q/2\}\) (Algesheimer et al., 2002). For small values of \(d = 2t + 1\) this guarantees small absolute values for the coefficients and saves computing time.

3 PRELIMINARY RESULTS

Tables 1 and 2 give the comparison between step 2 of the unmodified GRR protocol and the optimizations made in (Lory, 2009) and in this work, respectively. The first version is the straightforward implementation of the unoptimized GRR protocol, with coefficients \(\lambda_i\) in the interval \(\mathbb{Z}_q\); the second version is designed for small values of \(n\) as presented in (Lory, 2009); the third version finally exploits the observations of this work and uses coefficients \(\lambda_i\) from \(\mathbb{Z}_q\).

All the computations use the GNU multiple precision arithmetic library \(^1\) in version 5.0.1 and are on an AMD Athlon64 X2 5200+ with one physical core deactivated, fixed to 1.0GHz. The results obtained on this setup can obviously not be compared to those achievable on actual sensor hardware, but if the cycle count on this test setup is already far too large, the proposed solution obviously does not work as expected.

### Table 1: Comparison of the running time in milliseconds of step 2 of the unmodified GRR protocol and our optimizations of this protocol, as published in (Koschuch et al., 2010).

<table>
<thead>
<tr>
<th>(k)</th>
<th>(GR)</th>
<th>(Koschuch et al., 2010)</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>0.047</td>
<td>0.018</td>
</tr>
<tr>
<td>9</td>
<td>0.154</td>
<td>0.081</td>
</tr>
<tr>
<td>33</td>
<td>2.218</td>
<td>2.866</td>
</tr>
<tr>
<td>129</td>
<td>40.495</td>
<td>154.847</td>
</tr>
</tbody>
</table>

### Table 2: Comparison of the running time in milliseconds of step 2 of the unmodified GRR protocol and the additional optimizations of this protocol from this work.

<table>
<thead>
<tr>
<th>(k)</th>
<th>(GR)</th>
<th>Reduction to (\mathbb{Z}_q)</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>0.047</td>
<td>0.009</td>
</tr>
<tr>
<td>9</td>
<td>0.154</td>
<td>0.027</td>
</tr>
<tr>
<td>33</td>
<td>2.218</td>
<td>0.371</td>
</tr>
<tr>
<td>129</td>
<td>40.495</td>
<td>12.588</td>
</tr>
</tbody>
</table>

Tables 1 and 2 show the comparison of step 2 of the unmodified protocol with the optimizations detailed in (Lory, 2009) and the ones performed in this work, respectively. Our new approach with reduction to \(\mathbb{Z}_q\) improves the running times significantly, up to a factor of 6 when compared to an unmodified GRR implementation. In addition, it can be assumed that this reduction also results in significantly less memory requirements during protocol execution, although this still remains to be proven by complementary measurements.

4 OUTLOOK

Our preliminary results look promising and clearly indicate an additional performance improvement when using the optimizations proposed in this work. The

\(^1\)http://gmplib.org
next steps will be to replace the GMP library with our own code, optimized for constrained devices and much smaller than the GNU library and finally porting the protocol to a sensor node to get the timings on real hardware.

In addition, we also plan to perform a more detailed analysis of the algorithm, including several different bitlengths and numbers of players.

ACKNOWLEDGEMENTS

Manuel Koschuch, Matthias Hudler, and Michael Krüger are supported by the MA27 - EU-Strategie und Wirtschaftsentwicklung - in the course of the funding programme “Stiftungsprofessuren und Kompetenzteams für die Wiener Fachhochschulausbildungen”. Peter Lory is supported by the European Regional Development Fund - Europäischer Fonds für regionale Entwicklung (EFRE).

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