Performance Analysis of SyncML Server System Using Stochastic Petri Nets

Byung Yun Lee, Gil Haeng Lee, and Hoon Choi

I. Introduction

Synchronization is the process of making data consistent with data that is distributed via various devices such as PDAs, notebooks, and desktop computers. Device manufacturers have developed their own synchronization mechanisms, and as a result, there is a lack of interoperability between devices and application services. To solve these problems, many device manufacturers such as Nokia, Ericsson, IBM, and so on have prepared the Synchronization Markup Language (SyncML) protocol specification [1].

In this paper, we propose an enhanced architecture for the SyncML server system which has the capability of synchronizing personal information management service data. To improve the performance of the data synchronization server, we have designed and implemented a software frame, the elements of which are a sync adapter, sync agent, sync engine, session manager, and an open database interface based on its own processing functions. In this study, we present the enhanced architecture of our system based on the concept of a software frame. To design the system in a frame-based manner provides more portability and scalability. The sync agent has the role of processing the synchronization commands, which are independent of specific applications. The sync engine has the role of processing the actual synchronization in the system. The session manager maintains the session information between a client and server. The open database interface has the role of processing and adapting all kinds of database systems.

To evaluate our system, we model our system using stochastic Petri nets and evaluate the model using the stochastic Petri nets package (SPNP) 6.0 [2], [3], and we analyze various performance factors including throughput, buffer length,
system utilization, and response time. Lastly, we present conclusions regarding our research.

II. SyncML Server Architecture

The SyncML server in this study consists of eight frames, as shown in Fig. 1. The server application provides an interface for a user or the application administrator to modify application data at the server. The server adapter passes messages from client devices to the SyncML toolkit via the communication adapter [4]. The server adapter also passes parsed data from the toolkit to the sync agent. The SyncML toolkit encodes and decodes SyncML messages.

The sync agent implements the SyncML protocols. Therefore, its function is application independent and can be commonly used for all kinds of SyncML applications. The sync engine implements application-dependent parts such as service policies, as well as resolution rules in case of data conflicts. A brief scenario of the SyncML server is as follows.

i) The client wants to synchronize the data to the server.

ii) The client builds the SyncML message based on the data from the database. The client sends a set of commands which calls a function that sends a message to the server using the available transport binding.

iii) The sync agent in the server analyzes the received message and finds changed records.

iv) The sync agent finds the available device which must change and requests the sync engine to modify the records of all devices because it is necessary to create consistency between different devices of same user.

v) The sync engine makes and sends the status message to the client.

III. Performance Comparison

To evaluate the performance of our server, we measured the time to process one synchronization session in each type of server. The implementation and test environment was Microsoft Visual Studio C++ on Windows platforms, and the measurement was carried out with the parameters presented in Table 1.

As shown in Table 2, we measured the times from the SyncML demo architecture (SDA), SyncML Server 1.0 (session information in database), and Server 1.1 (session information in memory) under the same conditions. The SDA is the model architecture that was presented by the SyncML Initiative. The time we measured is the processing delay by the server only; it does not include the message transmission delay on the network. Test messages include mixed commands including add, replace, and delete commands.

Figure 2 depicts the results from the performance measurement. As we anticipated, SyncML Server 1.1 performs best. SyncML Server 1.1 is 44% faster than the SDA and 31% faster than SyncML Server 1.0. This is because the framework-based design and session-management mechanism in our system shows a better performance than that of existing servers based on the SDA. The Session Manager in SyncML Server 1.1 manages the session information of each device, and thus it has a faster processing time than the other mechanisms assessed here.
### Table 2. Comparison with existing server.

<table>
<thead>
<tr>
<th></th>
<th>Existing server (SDA2)</th>
<th>Our server</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Architecture</strong></td>
<td>Does not manage session (DLL structure)</td>
<td>Session is managed by session manager frame in a memory</td>
</tr>
<tr>
<td><strong>Database</strong></td>
<td>Dedicates DB (Lotus domino)</td>
<td>Provides open DB connectivity using open DB interface frame</td>
</tr>
<tr>
<td><strong>Portability</strong></td>
<td>Single structure without a frame</td>
<td>Divides by frame based on the processing functions, so enhances portability and scalability.</td>
</tr>
<tr>
<td><strong>Processing method</strong></td>
<td>Processes a set of commands without managing session, so it can process only one service.</td>
<td>Processes command with managing session, so it can support more than one service.</td>
</tr>
</tbody>
</table>

### IV. Petri Net Model

This section presents a Petri net model for the performance analysis of the proposed SyncML server system. Figure 3 illustrates the SyncML server system implemented for data synchronization. For a steady state evaluation, we use the SPNP which is a versatile modeling tool for the solution of stochastic Petri net (SPN) models [2], [5]. The SPN models are described in the input language for SPNP, C-based SPN language (CSPL), which is an extension of the C programming language with additional constructs that facilitate easy description of SPN models. SPNP can be employed to solve non-Markovian SPNs and is used to obtain the analytic numeric solution of Markovian model discrete-event simulations [6].

### V. Performance Evaluation of SyncML Server

To evaluate the performance of the implemented server, we use SPNP 6.0 which is based on a stochastic Petri net model [7], [8]. In this model, we define the arrival rate ($\lambda$), service processing rate of the sync adapter ($\mu$), memory size ($pM$), and number of concurrent DB connections ($pDB$). For each case, we change the value of $\lambda$ and analyze the average number of waiting jobs in the input buffer, the probability that the input buffer is empty, and the bottleneck frame. We also evaluate system response time [9].

1. **Measurement for the Parameters**

To decide the parameter values of the Petri net model, we measure the real processing time per command in each defined frame (Table 3), and we reflect these values in the proposed Petri net model (Table 4). This measurement was performed in Windows 2000 (CPU: Pentium 1.4 GHz, Main memory: 128 MB). We defined the processing time for each frame as follows.

2. **Processing Rate for the Defined Model**

From the measurement of the processing time, we can obtain the rate of the exponential distribution rate, taking the reciprocal number of the mean processing time.

3. **Performance Analysis**

#### A. Average System Throughput

Throughput is the term used for the amount of commands that the system can process in a unit time [10], [11]. Figure 4 presents the system throughput of our system.

As shown in Fig. 4, with an increase of $\lambda$, the average system throughput of the modeled system increases gradually until $\lambda$ is 1.5. When $\lambda$ is above 1.5, the throughput of the system decreases until $\lambda$ is 2.0, which is maintained as a constant value. This means that, in a given circumstance, the average system throughput of the modeled system is best when the arrival rate of the commands is near 1.5. These decreases in capacity are caused by a preemption of the system resources. As a result, we can estimate the best throughput of our server system before deploying the services.

#### B. Average Number of Waiting Jobs in the Input Buffer

The request messages from the client wait for processing in the server input buffer. If many jobs arrive in the input buffer, those beyond the number of $p1$ are discarded.

As shown in Fig. 5, with an increase of $\lambda$, the average number of waiting jobs in the buffer increases gradually until $\lambda$ is 1.5. When $\lambda$ is above 1.5, the average number of waiting jobs maintains a constant value. This means that the average number of waiting jobs is not in excess of the number of input buffers in a server; it increases gradually, and when it comes to reach the threshold of the system, it then converges on the predefined value.

#### C. The Probability the Input Buffer is Empty

If the job arrival rate ($\lambda$) is too low or the system processing rate is too high, the system input buffer is almost empty, and therefore it is profitable to reduce the input buffer size or degrade the hardware of the system. Figure 6 presents the probability that the input buffer is empty.

As shown in Fig 6, with an increase of $\lambda$, the probability that the input buffer is empty decreases gradually. This means that if the job arrival rate exceeds the processing capability of the system, the probability converges on 0.
Fig. 3. Petri net model for SyncML server.

Table 3. Average processing time per command (sec/command).

<table>
<thead>
<tr>
<th></th>
<th>Tadapter</th>
<th>Tsa</th>
<th>Tse</th>
<th>Todbi</th>
</tr>
</thead>
<tbody>
<tr>
<td>Add</td>
<td>0.17</td>
<td>0.22</td>
<td>0.612</td>
<td>0.12</td>
</tr>
<tr>
<td>Replace</td>
<td>0.17</td>
<td>0.1732</td>
<td>0.51</td>
<td>0.14</td>
</tr>
<tr>
<td>Delete</td>
<td>0.126</td>
<td>0.1071</td>
<td>0.245</td>
<td>0.1</td>
</tr>
</tbody>
</table>

Table 4. Number of commands per second (commands/sec).

<table>
<thead>
<tr>
<th></th>
<th>Tadapter</th>
<th>Tsa</th>
<th>Tse</th>
<th>Todbi</th>
</tr>
</thead>
<tbody>
<tr>
<td>Add</td>
<td>5.88</td>
<td>4.55</td>
<td>1.63</td>
<td>8.33</td>
</tr>
<tr>
<td>Replace</td>
<td>5.88</td>
<td>5.77</td>
<td>1.96</td>
<td>7.14</td>
</tr>
<tr>
<td>Delete</td>
<td>7.94</td>
<td>9.34</td>
<td>4.08</td>
<td>10</td>
</tr>
</tbody>
</table>

Tadapter: Average processing time in a sync adapter frame
Tsa: Average processing time in a sync agent frame
Tse: Average processing time in a sync engine frame
Todbi: Average processing time in an open DB interface frame
Fig. 4. System throughput ($p_0 = 10$ and $p_0=20$).

Fig. 5. Average number of waiting job.

Fig. 6. The probability that the input buffer is empty.

D. Bottleneck Frame

Our developed system consists of frames, and while processing the various synchronization commands, a bottleneck may occur in a specific frame. This may be influenced by the type of services and synchronization commands.

To analyze the bottleneck frame of the server, we evaluate the proposed Petri net model for the following cases ($p_0 = 10$, $\lambda = 2$, $\mu = 1$, $pM = 10$, $pDB = 1$) and count the average number of tokens in each place. As shown in Fig. 7, places $p1$ and $p4A$ have many tokens to process the synchronization. This is because all the synchronization requests from the client must arrive and wait in the input buffer $p1$, and $t4A$ takes the longest time to complete its operation. In this SPNP model, we define the reward function probability to reflect the distribution of synchronization commands. This reward function depends on the application of the synchronization; in this model, we assume that reward functions of the add commands constitute 70 percent of the total synchronization commands.

E. System Response Time

System response time is the processing time for requests from the client to arrive at the server, which processes the synchronization commands, and the response messages to return to the client. It is very important to forecast the response time of the designed system. We have measured the real processing time of the developed system, and we reflected the various factors of the system to the defined Petri net model. In doing so, we can change the various factors that affect the system performance and estimate the real performance of the system for many different cases such as the job arrival rate, processing rate of the input buffer, and system resources defined in the Petri net model. To measure the response time of the system, we redefine the Petri net model with respect to the
add command and measure the response time of the add command. We assume that there is a single token in p0. This is done in order to simplify the redefined model and minimize the states of the system.

\( \lambda \) is the arrive rate per second (job/sec). Thus if \( \lambda = 0.01 \), then 1 job arrives at the system in 100 seconds. Therefore, the inverse number of the arrival rate is the inter-arrival time between continuous jobs. We define the \( Ra \) as the response time of the add command and \( Pa \) as the probability that a token exists at p0. Then,

\[
Pa = \frac{1/\lambda}{1/\lambda + Ra}.
\]

From (1)

\[
Ra = \frac{1}{Pa} - 1.
\]

As defined in (2), the mean response time of add command \( Ra \) is determined by the job arrival rate, \( \lambda \), and the probability, \( Pa \). Using pr_std_average() in the SPNP package, we can obtain the probability of a token existing in p0A. The probability of the token in p0A (p0 = 5, \( \mu = 1 \), pM = 15, pDB = 1) is given in Table 5. We can also obtain the average response time using (2).

As shown in Table 5, using the configuration of a windows 2000-based personal system, we estimated that system response time \( Ra \) is about 1 second.

4. Performance Estimation

In the proposed model, if pM and job processing rate \( \mu \) are fixed, then the throughput of the system increases steadily until the system exhausts the resources; at which point, the throughput shows a slight decrease. As a result, we can estimate the optimal throughput for various situations.

As shown in Fig. 8, with an increase of pM, the throughput of the system can be obtained in various circumstances. In each case, we can estimate the best throughput of the system before deploying the server system.

VI. Conclusions

Estimating the throughput of a system is important to analyze and optimize the implemented system. The SPNP package used in this study is based on the Markov model, which represents the behavior of the system as a Petri net model, and transforms it into an analytic Markov model. Through this package, we can measure the various performance-related characteristics of the modeled system. We employ analytic modeling for the developed system in order to focus on the real factors that influence the system.

In this paper, we model the system exactly and adapt the various factors that are acquired from the developed system. We evaluate the processing time of synchronization commands in each frame of the system and define the reward function, adapting it to the Petri net model.

Using this model, we can evaluate various performance measures in different situations, and we can estimate the relationship between the arrival rate and the system resources.

As a result, we can estimate the optimal amount of resources due to the arrival rate before deploying the developed system.

References


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