Timed Automata Based Analysis of Embedded System Architectures

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Abstract

We show that timed automata can be used to model and to analyze timeliness properties of embedded system architectures. Using a case study inspired by industrial practice, we present in detail how a suitable timed automata model is composed. Exact upper bounds on the timeliness properties can be found with the Uppaal model checker for a number of usage scenarios. We compare our results with three other performance modeling techniques. This comparison shows that if the state space of the model is tractable, Uppaal gives the most accurate results at similar cost. The proposed modeling strategy can be automated, which alleviates the difficulty and error-proneness of manually constructing timed automata models.

1 Introduction

In the area of embedded systems, the market pressure to reduce the cost price of, for example, consumer electronics is very high. This requires the available resources in the system to be used to its fullest. It is often not economically viable any longer to oversize the design to compensate for unknowns in current and future performance requirements. This problem is enhanced by the fact that the demand for functionality is always increasing and the time-to-market targets for the production of these products are constantly decreasing. Therefore, designers need to shift their attention from the informal question “Does the product work?” to the question “Does the product work, given a set of hard resource restrictions?” This increases the importance of performance analysis in industry, in particular in the early phases of the system life cycle.

While in the past performance analysis was regarded as an activity that was primarily needed for the design of hard real-time systems, it is now recognized to apply to a much larger class of systems. Higher demands for usability aspects require the use of these techniques also in soft real-time systems, for example to analyze user interface responsiveness for gaming consoles or synchronization between audio and video stream decoding for DVD playback.

Our work is focused on performance evaluation of system designs in the early phases of the product life-cycle. This phase is characterized by its volatility, because there are still many unknowns that have great potential impact on the design. Paradoxically, it is also the life-cycle phase where the major architectural design decisions are taken. Obviously, performance analysis techniques are needed to support this decision making process. This support, however, can only be effective if it is very easy to construct, modify and analyze such a performance model. The turn-around time should be short to keep up with the interactivity of the design process.

In previous work [7], we explored the design of an in-car radio navigation system using real-time calculus. The aim of the experiment was to determine which proposed distributed embedded system architecture is best suited for the applications and their associated timeliness requirements. The results inspired us to explore other performance techniques using the same case study. In this paper we investigate timed automata.

Timed automata [1] are considered to be potentially useful for the analysis of systems, because of the expressiveness offered by this technique. Wide spread use in industry seems to be hampered by two issues: a) the general accepted belief that any realistically sized model will lead to a state space explosion during analysis and b) that despite its expressiveness, it remains hard to create timed automata models that reflect a particular design problem.

The AMETIST project has shown that significant progress has been made to tackle in particular the first question. Advances in tool development (such as symmetry reduction, efficient state representation and partial order evaluation among others) have enabled analysis of a number of very challenging industrial case studies, see http://ametist.cs.utwente.nl. For the second problem, two solution strategies are proposed in general. First, the timed
automaton framework itself can be extended such that it provides concepts that are closer to the problem at hand. Second, another (more domain specific) language can be used as a front-end; a timed automata model should then be automatically derived from such a specification.

In this paper we show that the second solution is indeed possible. We derive timed automata models from UML sequence diagrams that are augmented with performance data. Together with the increased performance of the associated analysis tools, we believe that this improves the use of timed automata in practice.

Related work. Petriu and Woodside [6] have shown that performance models can be generated from a set of scenarios and their use of resources. They take the UML profile for Schedulability, Performance and Time as their basis. Hence, we have not attempted to automate the model construction process ourselves because we expect that in due course commercial products will become available. Bennett, Field and Woodside [3] follow a similar approach to what we propose in this paper, but then using the stochastic process algebra FSP instead of timed automata (and LTSA instead of Uppaal respectively). They use a combination of model checking for verifying functional correctness properties and discrete event simulation for investigating the performance properties of the system, while we stay within one technique.

2 The case study

Figure 1 presents an overview of the in-car radio navigation system. The diagram suggests that there are three processors in the system interconnected by a communication bus. Each processor is assigned its own cluster of functionality, of which there are three: the man-machine interface (MMI), the navigation functionality (NAV) and the radio functionality (RAD).

We will consider three independent applications that run on the system concurrently: ChangeVolume, AddressLookup and HandleTMC. We will investigate whether the combination ChangeVolume and HandleTMC meet the system-level timeliness requirements for this architecture. First, we will characterize each application by presenting a UML sequence diagram that is augmented with performance data. In combination with the deployment and performance data shown in Figure 1, it provides sufficient information for modeling and analysis, which will be presented in Sections 3 and 4.

Figure 2 represents the typical use of the volume control of the radio. Note that three operations are identified: HandleKeyPress, AdjustVolume and UpdateScreen. Worst-case execution times, event rates and message sizes are estimated and annotated in the sequence diagram, together with the principle timing requirements (K2V and A2V) applicable to this scenario.

The TMC message handling scenario is shown in Figure 3. Note that the radio only receives the TMC messages; decoding is performed on the navigation subsystem because the map database is needed for that. Only fully decoded and relevant messages are presented to the user.

In this paper we analyze the combination of the
“ChangeVolume” and “HandleTMC” scenarios and the combination of the “AddressLookup” and “HandleTMC” scenarios. Since the modeling of the scenarios is very similar, we decided to omit the “AddressLookup” scenario here for brevity and only mention the analysis results of each combination. The interested reader can find a detailed informal description of all scenarios on-line at http://www.cs.ru.nl/ita/publications/papers/martijnh. Of course, in practice many more applications might be running in parallel to the three considered here. Note that the approach presented in the next section is not theoretically limited in the number of applications that can be analysed.

3 Modeling the system

The Uppaal model checker, which we use in this paper, is a tool for modeling and verifying networks of timed automata [2]. We assume that the reader has some knowledge about timed automata and their semantics. We will highlight where we use Uppaal specific language features. However, we do not describe the complete model\(^1\), we merely present the patterns used for modeling.

First, we will provide an informal intuition to the modeling approach. We represent each hardware component as a separate timed automaton. The hardware component is either idle or busy computing some function. Similarly, each communication link is modeled as a timed automaton. Each link is either idle or busy transporting some data.

A system can then be represented by the network of timed automata composed of the automata representing the hardware and the automata representing the communication links. As a consequence, for each deployment diagram we need to construct a separate network of timed automata, because the functionality is distributed differently over the given hardware.

The external workload of the system is also characterized using timed automata; events are generated and inserted into the system according to the workload specification. The response of the system is observed by the same automaton such that the timeliness requirements can be expressed and verified by the model checker.

We will give an example of a hardware component in Section 3.1. Then, we will show how the communication is modeled in Section 3.2 and finally, we will present the automata representing the environment in which the system operates in Section 3.3.

3.1 Modeling the hardware

Figure 4 presents the basic automaton that models the behavior of the radio functionality (RAD). From the two sequence diagrams (Figures 2 and 3), it can be deduced that this functionality in fact consists of two operations, AdjustVolume and HandleTMC respectively. Each operation is represented as a location in the automaton. The automaton has a local clock \(x\) and two local constants, HTMC and AV. These constants represent the execution time of the operation, which is simply calculated as the worst-case execution time (expressed by the number of instructions to execute, as specified in the applicable sequence diagram) divided by the capacity of the hardware component (which is expressed in million instructions per second, as specified in Figure 1).

![Figure 4. The automaton RAD](image)

This calculation is obviously an approximation that in general does not hold. The peak capacity of modern CPU architectures can only be achieved for specific types of algorithms; digital signal processors are optimized for repeated multiply-addition operations that occur frequently in fast Fourier transformations for example. But despite this approximation, it suffices for our purpose since we are interested in high-level analysis of early design models. If we want to make a more accurate prediction, it is always possible to benchmark the operation on the target architecture and use the measured value instead.

The global variables rec and setvolume keep track of the number of pending calls to HandleTMC and AdjustVolume respectively. The transition is enabled if either counter is greater than zero. The hurry! synchronization is so-called urgent and always available. It thus enforces greedy behavior: transitions with this synchronization must be taken as soon as they are enabled. This ensures that the model does not delay when there still are outstanding requests.

Thus, if the RAD automaton is in location idle and the Receive event arrives (which is modeled by the increment of the rec variable), then it immediately takes the transition to the location handleTMC (“immediately” means that no time elapses between the arrival of the event and the execution of the transition). The event is removed from the input queue (by decrementing the global variable rec) and the clock \(x\) is reset. The automaton stays for HTMC time units in location handleTMC and then returns to the idle location while generating an output event (receive_out is incremented). As we will see in Section 3.2, the global variables rec, setvolume, receive_out and getvolume_out are the interface to the communication link. Similarly, we will see in Section 3.3 that the synchronization no-
tice audible_change! is used to emit the completion of the AdjustVolume operation back to the environment for measuring the response time.

Note that the RAD automaton in Figure 4 models a non-deterministic non-preemptive scheduler, which is in most cases not very realistic. The Uppaal language allows to model many kinds of schedulers. For instance, the automaton in Figure 5 models the radio functionality again, but now with a fixed priority preemptive scheduling strategy, in which the AdjustVolume operation has priority over the HandleTMC operation.

In Figure 5, modeling priorities is achieved by the additional guard expression setvolume == 0 to the guard of the transition from location idle to location handle_TMC. This means that TMC messages may only be handled if there are no outstanding ChangeVolume requests pending. In order to model preemption, an additional local clock y and a local integer variable D are introduced to the automaton. The transition from idle to handle_TMC sets D to HTMC. Whenever a AdjustVolume event arrives during the execution of HandleTMC, the transition to hdl_pre is taken. Clock y is then used to measure the execution time of the ChangeVolume call that interrupted the execution of HandleTMC. When ChangeVolume is completed, the variable D is incremented with the delay caused by the preemption. Thus, preemption can easily be modeled, but it is crucial that it is known in advance how long a task can be preempted. Furthermore, it must not be the case that a task can infinitely often be preempted since then D grows to infinity and model checking is not possible anymore. The model checker can be used to prove that this is not the case by verifying some finite upper bound of D (care has to be taken because an integer variable in Uppaal has a finite domain by definition). The modeling of the other hardware components follow the same pattern as described above and are therefore not depicted here.

3.2 Modeling the communication

Modeling the communication links in the system is surprisingly similar to the models we have presented for the hardware components. For each link, we create a separate timed automaton. A location is created in the automaton for each message that is sent between the hardware components that communicate through the same communication link. The location reflects the fact that the message is being sent.

The location is occupied for as long as the message transfer takes. In Figure 6, we use the constants BYTES4 and BYTES64 to represent the time to transfer 4 and 64 bytes respectively over the communication link. This constant is again simply calculated as the message length (in bits, which is specified in the augmented sequence diagrams) divided by the baud rate (which is specified in the deployment diagram). Obviously, this formula can be adapted to compensate for expected protocol overhead.

The hardware components interface to the communication link using (sharing) the global variables that count the number of outstanding messages of a particular type that need to be transferred over the link. If the SetVolume message from Figure 2 is to be sent from the MMI to RAD, then the MMI automaton will simply increment the global setvolume_out variable. If the bus is idle (and all other global variables are zero), then the transition towards the location sending_setvol location is immediately taken (again due to the hurry! synchronization); this will decrement the setvolume_out variable. The location sending_setvol is occupied for BYTES4 time units and then the transition back to idle is taken; this will in turn increment the global variable setvolume which in turn will enable the corresponding transition in the RAD automaton, as presented in the previous paragraph.

Note that we can again use the same strategy for dealing with priorities. The communication link in Figure 6 is very simple, if more than one message is available then there will be non-deterministic choice who will get the turn to claim the bus. The bus is then blocked for as long as the message transfer is in progress. This resembles many of the simple industrial serial bus interfaces such as RS-485 for example. Analogous to Section 3.1, we can implement priorities by adding additional guards to the transitions that point outward from the idle location. It is relatively easy to mimic priority based protocols such as the Controller Area Network (CAN) or time-triggered protocols. For example a solution for a TDMA bus concept is proposed by Perathoner et al in [5]. Less trivial however is the encoding of protocols that break large messages into pieces to prevent starvation. However, the approach presented has another interesting characteristic. If the interface (the global variables) remains the same, then it would be simple to replace a certain bus concept by another by merely replacing the bus au-
tomata. It would not affect the hardware components at all. We can then easily investigate the impact of different bus protocols for a given deployment.

3.3 Modeling the environment

There are two actors that exercise the system from the environment: (i) a user who initiates the ChangeVolume scenario and (ii) a radiostation that initiates the HandleTMC scenario. There are two timed automata for each actor: a “normal” automaton and a “measuring” automaton. Which one is included in the system (the network of timed automata we analyze using Uppaal) depends on the worst case execution time we want to investigate. If we want to measure the response time of the ChangeVolume scenario, we add the “measuring” automaton for the user and the “normal” automaton for the radiostation. Vice versa, if we want to measure the response time of the HandleTMC scenario, then we add the “measuring” automaton for the radiostation and the “normal” automaton for the user.

We consider four basic kinds of event arrival models: (i) periodic, (ii) sporadic, (iii) periodic with jitter and (iv) bursty event streams. For the strict periodic event model, an offset can be specified to force a phase shift in the signal (start of the first period). The periodic and sporadic event models can be expressed by automata as shown in Figure 7 (a-c). The event model for periodic behavior with jitter (where the jitter must be smaller than or equal to the period) can elegantly be expressed by the model proposed by Perathoner et al in [5], which is shown in Figure 7 (d).

The behavior of events is called bursty when the jitter becomes larger than the period of the event. This can be modeled, but it is much more involved than the previous model of periodic behavior with small jitter. The reason is that the subsequent intervals in which an event can occur now overlap. Figure 8 shows the Uppaal model for bursty behavior with period $P$, jitter $J$ and a minimal separation time between events of $D$.

This automaton has two local variables, pending and snd. The variable pending is incremented every $P$ time units (using clock $x$) which models that an event may be sent. The clock $y$ is used to keep track of the next deadline for sending an event. If an event may be sent (i.e., $z > D$ which models the inter-event separation time and pending > 0), then the event is sent and snd is incremented. This signals that the deadline for the next event may be incremented by $P$ time units. This is modeled by the reset of clock $y$ after $J$ (for the first event) or after $P$ (for the other events) time units. Note that this automaton is not very nice for the state space of the model: it has three local clocks (but if $D = 0$ then $z$ can be left out) and two local integer variables that both can count up to $J + 1$.

Every event automaton also has a “measuring” variant which is used to record the WCRT of a generated event.
These automata seem complicated at first sight, but are in fact very logical in structure. All measuring variants can therefore be automatically generated.

Figure 8. environment automaton describing event bursts

Figure 9. The measuring variant “rstat-m”

The automaton in Figure 9 is the measuring variant of the automaton of Figure 7(c). It has two additional integer variables, \( m \) (initially -1) and \( n \) (initially 0) and an additional clock variable \( y \). This automaton can generate input events in the same way as the automaton in Figure 7(c) by using the upper transition from \( L0 \) to \( L1 \) and the upper-left and lower self-loop of \( L1 \). The variable \( n \) keeps track of the number of events that have been fed into the system and for which no response (a synchronization over the notice_visible_change3 channel in this case) has been received yet. The automaton can also arbitrarily choose an event to measure the worst case response time (WCRT) by the lower transition from \( L0 \) to \( L1 \) and the upper-right self-loop of \( L1 \). (It is assumed that all queues are FIFO queues, i.e., events do not overtake each other, they are order preserving and that events are never dropped.) On those transitions \( y \) is reset to 0 and \( m \) is reset to the value of \( n \). Hence, \( m \) counts the number of responses that need to be seen before the one is een that is used for the measurement. Note that at most one measurement can be in progress and that \( m = -1 \) if no measurement is in progress. If \( m = 0 \) then the response of the event that is being measured is expected. If the response occurs, then the transition to \( seen \) is taken. This is the so-called “committed” location in which no time may elapse. The WCRT of the HandleTMC scenario can thus be found using Uppaal by finding the smallest \( C \) (using a simple binary search strategy manually) such that Property 1 is satisfied.

\[
AG(rstat-m.seen \rightarrow rstat-m.y < C)
\] (1)

The modeling of the other actors follows the same pattern as explained above and are therefore not depicted here.

4 Analysis of the Uppaal model

The Uppaal model has been analyzed for 5 different requirements and 5 different sets of environment models. The results have been collected in Table 1. (Uppaal version 3.6-alpha-1 has been used with default options.) The rows indicate which requirement is measured and in what context (combination of applications) the analysis was performed.

The first column shows the results for strictly periodic event streams with a user-defined offset of 0 for all events (\( po, F = 0 \), which reflects fully dependent (synchronous) environment models. In the second column, the results are shown for strictly periodic event streams with an unknown offset for all events (\( pno \), which reflects fully independent (asynchronous) environment models. The third column presents the analysis results for sporadic event streams (\( sp \)) where only a lower bound is specified for the event inter arrival time. In the fourth column we show periodic event streams with small jitter (\( pj, J = P \)) for the “radio station” environment model and sporadic events for the others. And finally, we use bursty event streams (\( bur, J = 2P, D = 0 \)) for the “radio station” and sporadic event streams for the others in the last column.

Analysis of the HandleTMC and AddressLookup scenarios proved to be no problem. The verification times for \( po, pno \) and \( sp \) where so small (typically less than a second) that a binary search could easily be performed. The \( pj \) and \( bur \) scenarios took a bit more effort, but still a binary search was feasible (verification times typically in the order of a few minutes). The HandleTMC scenario in combination with ChangeVolume proved to be a problem for the \( pj \) and
Table 1. Uppaal worst-case response time analysis results (in milliseconds)

<table>
<thead>
<tr>
<th>Requirement</th>
<th>Event model</th>
<th>po (F = 0)</th>
<th>pno</th>
<th>sp</th>
<th>pj (J = P)</th>
<th>bur (J = 2P, D = 0)</th>
</tr>
</thead>
<tbody>
<tr>
<td>HandleTMC (+ ChangeVolume)</td>
<td>357.133</td>
<td>381.632</td>
<td>382.076</td>
<td>&gt; 400.000 (df)</td>
<td>&gt; 500.000 (df)</td>
<td></td>
</tr>
<tr>
<td>HandleTMC (+ AddressLookup)</td>
<td>172.106</td>
<td>239.080</td>
<td>239.080</td>
<td>329.989</td>
<td>420.898</td>
<td></td>
</tr>
<tr>
<td>K2A (ChangeVolume + HandleTMC)</td>
<td>27.716</td>
<td>27.716</td>
<td>27.716</td>
<td>&gt; 27.715 (bf)</td>
<td>&gt; 27.715 (bf)</td>
<td></td>
</tr>
<tr>
<td>A2V (ChangeVolume + HandleTMC)</td>
<td>41.796</td>
<td>41.796</td>
<td>41.796</td>
<td>&gt; 41.795 (bf)</td>
<td>&gt; 41.795 (bf)</td>
<td></td>
</tr>
<tr>
<td>AddressLookup (+ HandleTMC)</td>
<td>79.075</td>
<td>79.075</td>
<td>79.075</td>
<td>79.075</td>
<td>79.075</td>
<td></td>
</tr>
</tbody>
</table>

Note that the AddressLookup and ChangeVolume worst-case response time values remain constant since (i) they have priority over the “radiostation” related events and (ii) their event model parameters are such that events are never queued for processing. For example, each AddressLookup event is fully processed before the next event arrives. If we would allow jitter to the AddressLookup scenario, such that two events might overlap, then the bound of 79.075 would increase.

5 Comparison with other techniques

In our earlier work [7], we have computed the worst-case response time results for several different distributed architectures using Modular Performance Analysis, but in this paper we only performed the analysis and comparison of the architecture shown in Figure 1. The main reason for this is that the timed automata models are constructed by hand which is a slow and very error prone process. We do believe that it is possible to automate the construction of these models, which is future work.

We did compare our analysis results with a few other techniques that were applied to the same case study. The results are summarised in Table 2. For space reasons it is not possible to discuss any of these models in detail. Note that the results for the other techniques where provided to us by other researchers that can be considered expert users for each respective tool.

The parallel object-oriented specification language (POOSL, http://www.es.ele.tue.nl/poosl/) and its SHESIM support tool was used to build a discrete event simulation model, which is typically used in industry nowadays. The simulation run of the periodic case with an unknown offset (pno) shows that the worst-case response time is not necessarily found by simulation, which is due to the fact that infinitely many values can be chosen for the offset parameter. Moreover, simulation runs typically took minutes to several hours.

Symbolic timing analysis for systems (SymTA/S, http://www.symtavision.com) is a performance and timing analysis tool based on formal scheduling analysis techniques and symbolic simulation. The tool supports heterogeneous architectures, complex task dependencies, context aware analysis, and combines optimization algorithms with system sensitivity analysis for rapid design space exploration [4]. The results found by SymTA/S were on par with the results found by Uppaal, with the exception of the periodic event streams with zero offset. In-depth discussions with the SymTA/S experts learned us that this type of event streams can in fact be handled but the specific technique needed is not yet fully implemented in SymTA/S. In stead, the pno results are used, which provides a safe upper bound for the WCRT.

Modular performance analysis is a deterministic queuing theory based on real-time calculus (MPA, http://www.mpa.ethz.ch). Again, the results found with MPA are quite similar to Uppaal and SymTA/S. The results are slightly more conservative mainly because context information is lost due to the transformation from the time to the time interval domain which is used by real-time calculus. The phase shift disappears when this transformation is performed. Therefore, it is not possible to get tight bounds for the po case with MPA, it will always yield the more conservative pno results.
Table 2. Worst-case response time results – comparison with other tools

<table>
<thead>
<tr>
<th>Requirement</th>
<th>Tool</th>
<th>Uppaal (pno)</th>
<th>POOSL (pno)</th>
<th>SymTA/S (pno)</th>
<th>MPA (pno)</th>
</tr>
</thead>
<tbody>
<tr>
<td>HandleTMC (+ ChangeVolume)</td>
<td>357.133</td>
<td>381.632</td>
<td>266.94</td>
<td>382.086</td>
<td>390.0862</td>
</tr>
<tr>
<td>HandleTMC (+ AddressLookup)</td>
<td>172.106</td>
<td>239.080</td>
<td>244.26</td>
<td>253.304</td>
<td>265.8491</td>
</tr>
<tr>
<td>K2A (ChangeVolume + HandleTMC)</td>
<td>27.716</td>
<td>27.716</td>
<td>27.7067</td>
<td>27.717</td>
<td>28.1616</td>
</tr>
<tr>
<td>A2V (ChangeVolume + HandleTMC)</td>
<td>41.796</td>
<td>41.796</td>
<td>41.7771</td>
<td>41.798</td>
<td>42.2424</td>
</tr>
<tr>
<td>AddressLookup (+ HandleTMC)</td>
<td>79.075</td>
<td>79.075</td>
<td>78.8989</td>
<td>79.076</td>
<td>84.066</td>
</tr>
</tbody>
</table>

6 Conclusion

We have shown that timed automata can be used to analyze timeliness properties of embedded system architectures. The proposed modeling approach is straightforward and can be automated. The benefit is that if the state space explosion problem can be controlled then exact and hard upper bounds are found during analysis, while this cannot in general be guaranteed for other techniques. Table 2 shows that the results found by our approach are indeed competitive to others. Some specific conclusions can be drawn from our experiments:

Uppaal is a very generic tool. Many model types and scenarios can be expressed (i.e. event models, communication, computation, mix of preemptive and non-preemptive elements) and analyzed. However, manual construction and maintenance of these models is error prone and should be automated to be useful in an industrial setting.

When the time scales of the environment models do not differ too much, then Uppaal is quite suitable for case-studies of this size. Of course, model checking still suffers from the state space explosion problem. Nevertheless it can also be used as a “massive simulation” tool: let the model checker search (e.g. random-depth-first) for any trace that gives a lower bound on the response time.

The comparison shows that the results fit the theoretical picture; the simulation-based approaches give values that are smaller than those computed by Uppaal (the worst-case instance is not necessarily reached) while SymTA/S and MPA provide values that are larger than those computed by Uppaal (hard but not necessarily tight values are computed).

Uppaal was able to find more accurate results with similar modeling and analysis effort for most scenarios of this case study. In fact, some results found by simulation could be falsified by showing the counter example from the model checker while searching for that particular outcome. Currently, however, Uppaal lacks the features that are necessary to conveniently perform a parameter sweep; something that MPA and SymTA/S are capable of.

Comparing the various tools has not been easy. The key difficulty has been to ensure that all models have the same semantics and deal with modeling assumptions in a similar way. Using the Uppaal verification engine, these assumptions could be checked mechanically which helped to improve all specifications.

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References