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High-Level Modeling of Communication-Centric Applications: Extensions to a System-Level Design and Virtual Prototyping Tool

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Abstract

High performance streaming applications require hardware platforms featuring complex, multi-level interconnects. These applications often resemble a task-farm, where many identical tasks listen to the same input channel. Usual embedded system design tools are not well adapted to capture these applications. In particular, the non-uniform memory access (NUMA) nature of the platforms induces latencies that must be carefully examined. The paper proposes a multi-level modeling methodology and tools (TTool, SoCLib) that have been extended to model the characteristics of streaming applications (multiple tasks, non deterministic behavior, I/O devices) in UML/SysML, and to automatically generate a virtual prototype that can be simulated with high precision. The paper uses a typical streaming application to show how latencies can be estimated and fed back to diagrams.

Keywords: Virtual prototyping, System-level design, Design space exploration

1. Introduction

Parallel telecommunication and video streaming applications often adhere to the \textit{task-farm paradigm}: they contain a high number of tasks, each waiting to accept data for processing from one of the common (input) buffers. Streaming applications also have the particularity that they require specific Input/Output (I/O) co-processors which read from, for example, an Ethernet link and, after processing by the application, write to another link or display video streams on a display.

\textit{Multiprocessor-on-chip system} (MPSOC) architectures feature complex, sometimes hierarchical, interconnect-on-chip architectures: they are thus well adapted for the deployment of highly parallel applications. These MPSOC usually rely on Non Uniform Memory Access (NUMA), where clusters using local crossbars are
grouped around a central interconnect. In these architectures, memory access
latencies are hard to predict, thus resulting in non-deterministic behaviors.

Design space explorations can help predicting application timings. Ideally,
the design space exploration should be performed at a high level of abstraction
so as to minimize the design effort to dimension the application and the under-
lying architecture. However, since high-level models are highly abstracted, it
means that decisions are taken with high-level parameters that should be verified
later in the system design process using more precise modeling and simulation
schemes (e.g. cycle accurate simulations).

Tool support is certainly required to carry out this method. Yet, in the
field of classical embedded systems for which many high-level modeling and
exploration tool exists, communications are typically one-to one and do not
fully exploit the capacities of such architectures e.g. the standard interconnect
is a (simple) bus.

Our previous work [1, 2] enabled the multi-level modeling of high throughput
task-farm applications, by providing a formal basis for non-deterministic chan-
nel accesses and extended TTool to model NUMA architectures and generate
NUMA virtual prototypes. However, while streaming applications strongly rely
on I/O elements, it was not possible to model these elements. Neither could
we explicitly take into account the very strong variations of latencies, e.g. for
memory access, in the model.

In summary, our contribution presents an overview of new and existing
work on a model-based engineering (MBE) approach for complex task-farm type
streaming applications with a presentation of the overall method, the semantics,
modeling extensions, as well as a comprehensive case study.

The paper is organized as follows. Section 2 presents the related work.
Section 3 presents the background frameworks. Section 4 explains extensions
of the semantics to capture non-determinism, and of the hardware model to
convene for our class of applications. Section 5 illustrates the approach by
means of a case study, for which experimental results are presented in Section
6. Section 7 concludes the paper and gives our perspectives.

2. Related Work

During an embedded software development process, software components are
generally tested/executed on a local host, and then are integrated once the target
is available. The lack of information regarding software/hardware interactions
often leads to late and costly software revisions. Moreover, reconsideration
of software design choices may strongly impact MPSoC design, because they
feature a high number of processor cores and complex interconnection networks.
Thus, we require a fast and easy-to-use solution for a preliminary high-level
exploration.

One possible approach is to frequently validate the different refinements of
software components in a as-realistic-as-possible hardware environment. Flexi-
ble Programmable Gate Arrays (FPGAs) can be used for this purpose [3, 4], but

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the hardware elements must be totally developed to include a sufficient amount of details, and then flashed onto a —sometimes costly— FPGA.

A less expensive solution is the use of virtual prototyping platforms. They are slower and less realistic than FPGAs, but offer more flexibility, and are more accessible for software engineers with limited hardware knowledge.

Many completely software-based prototyping environments have been proposed. Some of them are restricted to high-level analysis and offer only functional simulation, while others offer virtual prototyping and extended profiling capabilities.

PtolemyII [5] proposes a modeling environment for the integration of diverse execution models, in particular hardware and software components. Even if design space exploration can be performed, its first intent is the simulation of the modeled systems.

In Polis [6], applications are described as a network of state machines. Each element of the network can be mapped on a hardware or a software node. This approach is more oriented towards application modeling, even if hardware components are closely associated to the mapping process. Metropolis [7], an extension of Polis, targets heterogeneous systems and offers various execution models. Architectural and application constraints are however closely interwoven.

Sesame [8] proposes modeling and simulation at several abstraction levels. In contrast to Metropolis, application and architecture are clearly separated in the modeling process. Models’ semantics vary according to the levels of abstraction, ranging from Kahn process networks (KPN [9]) to data flow for model refinement, and to discrete events for simulation. Currently, Sesame is limited to the allocation of processing resources to application processes. It neither models memory mapping nor the choice of the communication architecture.

The ARTEMIS [10] project originates from heterogeneous platforms in the context of research on multimedia applications in particular, thus justifying the acronym (ARchitecTurEs and Methods for embedded MedIa Systems); it is strongly based on the Y-chart approach. Application and architecture are clearly separated: the application produces an event trace at simulation time, which is read in by the architecture model. However, behavior depending on timers and interrupts cannot be taken into account.

MARTE [11] shares many commonalities with our approach, in terms of the capacity to separately model communications from the pair application-architecture. However, it intrinsically lacks separation between control aspects and message exchanges. Other works based on UML/MARTE, such as Gaspard2 [12], are dedicated to both hardware and software synthesis, relying on a refinement process based on user interaction to progressively lower the level of abstraction of input models. Still, such a refinement does not completely separate the application (software synthesis) or architecture (hardware synthesis) models from communication. Finally, MARTE neither offers explicit support for task-farm applications nor for NoC and NUMA based platforms.

Di Natale et al. [13] propose the generation of communication managers for software low layers. Yet, they do not handle the specificity of task-farm applications nor do they offer formal verification.
Batori [14] proposes a design methodology specific for telecommunication applications. From use cases, the method proposes several formalisms to capture the application structure (“interaction model”) and behavior (Finite State Machine) and for its deployment from which executable code can be generated. The platform seems limited to specific components and no design exploration seems possible; code generation targets a real platform, and not a prototyping environment.

An important aspect of streaming applications is I/O, requiring hardware modules for the sending and receiving of network packets or video frames either from a network, a file or other source. Often very specific, they are difficult to model in tools for modeling generic applications [15, 16].

We therefore propose an approach to address all aspects of such applications, in terms of software and hardware modeling, formal verification, simulation, and performance evaluation, using a model-driven approach and existing toolkit as a basis.

3. Background

3.1. Supporting Toolkit

TTool [17] is a UML/SysML modeling and verification toolkit for designing embedded systems, supporting different level of abstraction, and proposing a push button approach for both simulation and formal verification.

3.1.1. Modeling Levels

The main abstraction levels of TTool are partitioning and software design [18]. The software design level includes a prototyping phase in which software can be simulated onto the hardware using a precise simulation engine.

As shown in Figure 1, the approach is as follows:

1. The overall method starts by a partitioning phase, DIPLODOCUS (Design sPace expLoration based on fOrmal Description teChniques, Uml and SystemC), containing three sub-phases: the modeling of the functions to be realized by the system (functional view), the modeling of the candidate architecture as an assembly of highly abstracted hardware nodes, and the mapping phase. A function mapped on a processor is a software function, a function mapped on a hardware accelerator corresponds to a custom ASIC (Application-specific Integrated Circuit).

2. The second phase is based on AVATAR (Automated Verification of reAl Time softwARe [19]), a SysML-based environment for modeling the software components of complex embedded systems. It starts by the design of the software and hardware. A deployment view shows the allocation of software components. Code can then be generated both for the software components of the application (in C/POSIX code) and for the virtual hardware nodes (in System C format).
Choice of parameters at the higher level of abstraction is subject to validation or invalidation due to experimental results on the generated prototype. Thus, simulation results at the prototyping level could lead to reconsideration of the partitioning decisions.

At each stage, simulation and formal verification ensure that our design meets performance, behavior, and schedulability requirements [20]. Simulations and verifications first rely on model transformation techniques e.g. SysML-to-formal specification. Safety properties can be evaluated using the TTool internal model checker or with UPPAAL [21], while security properties can be proved with ProVerif [22].

3.1.2. Simulation

From partitioning models, high-level simulation can be performed. From software design models, an application and its deployment can be transformed into a virtual prototype [23] based on SoCLib [24], a public domain library of component models written in SystemC. SoCLib targets shared-memory MPSoC architectures based on the Virtual Component Interconnect (VCI) protocol [25]. This protocol clearly separates the components functionality from communication aspects; we use Cycle/Bit Accurate component models.

Basically, the transformation works as follows. Each SysML block is translated into a POSIX thread and executed on one of the general purpose processors. The threads uses on primitives defined in the AVATAR runtime. These primitives capture the semantics of the AVATAR operators (delay, asynchronous read, etc.) and implement them using C/POSIX calls. The main program instantiates the POSIX threads of the AVATAR blocks, and the channels translated as software objects stored in the on-chip memory. Threads are spawned from the main thread on the CPU indicated in the deployment diagram. The
**top cell generator** generates a SystemC top cell from the deployment diagram. A **ldscript generator** generates the linker script taking into account the mapping specified in the deployment diagram.

### 3.1.3. Capturing non deterministic behaviour

In [1], we have shown how to rely on non-deterministic operators of timed automata in order to capture **multi-writer multi-reader (MWMR)** communications as already explained in in [26] i.e. any number of reader or writer tasks can access simultaneously to these channels. In other words, a task waiting for some data to be processed indifferently picks up data from common buffers, thus introducing a high degree of non deterministic behaviors that are difficult to capture in FIFO models (e.g., Kahn models). However, at the time, these channels lacked formal semantics, and generation of platform variants was text-based and semi-automatic. Due to our contributions, it is now possible to work from high-level models, generate the prototyping code and feed the results back to the higher-level diagrams.

An example of non deterministic behavior —using the *after* clause of SysML state machines— is shown on the left side of Figure 2 where two SysML blocks communicate via a channel. Asynchronous channels can be rather naturally implemented using multi-writer multi-reader channels. In the figure, AVATAR asynchronous channels are depicted with origin and termination ports filled in white. The right part of the figure shows the state machine diagrams of blocks B0 and B1, respectively. This is a non deterministic system since e.g. B0 can decide to send either m0 or m1. B1 waits for one of the two messages, or if a given time without receiving one of the two (“after(10,20)”) passes, it resets the waiting time of messages by taking the top right transition from its main state.

![Figure 2: Non-deterministic AVATAR model](image)

### 3.1.4. Modeling task-farm applications in TTool

In the field of embedded systems, communications are typically one-to one, more rarely one-to-many e.g. broadcast communication. Many-to-many communications are not common practice. **Block hierarchy** can be used to express the fact that several blocks write to the same channel, as shown in previous work [1].
Figure 3 shows a typical task-farm application modeled as AVATAR block diagram. Masters and Workers are connected through an AVATAR channel regrouping all communication between the master and worker tasks (i.e. the common channel). The master blocks have to be put into a higher-level block called Masters, writing indifferently to the channels, and the worker into a higher-level block Workers, reading the channels, respectively. Figure 4 does not show the state machines of the inner tasks which are very simple, but only the two outer tasks. The Masters task reads data produced by either one of the two inner tasks Master0 and Master1 via channel from_master0 and from_master1, the Workers task dispatches data to either one of the two inner tasks Worker0 and Worker1 via channel from_worker0 and from_worker1.

3.2. Support for NUMA Architectures

TTool, up to the work described in [2], did not provide support for modeling NUMA virtual platforms. A NUMA architecture typically addresses at least two computing domains. TTool has been extended to handle NUMA architectures thanks to the following contributions:

• Modeling aspects. A local crossbar component and the possibility to connect communication components (crossbar, bus, mesh, ...) with each other to build multi-level interconnects have been added. TTool is still limited to two-level interconnects featuring local crossbars around either a Virtual Generic Multi Network (VGMN), abstraction of a $n \times m$ mesh interconnect, or a Virtual Generic Serial Bus (VGSB), abstraction of a serial bus.
• **Model transformation.** Automated code generation for NUMA architectures has been added. While the model-to-code translation of software tasks and the generation of a main program remains mostly the same with respect to what was described in [23], the *topcell* (in particular the determination of the segment’s addresses in the mapping table) as well as the *ldscript* are far more complex. Using a two-level interconnect in the shared memory paradigm means that addresses and segment sizes, which are automatically generated by the tool, must avoid overlapping and respect cluster rules (in Most Significant Bits) and local address; local and shared memory segments have to be generated. Each cluster contains at least one RAM. Additional infrastructures which are invisible from the SysML level are replicated on the clusters, such as DMA and interrupt controllers.

3.3. Capturing Latencies

NUMA platforms usually have a high access latency variance, because time required for memory access differs depending whether data is located on the same or a different cluster. To better study timings, our idea is to rely on multi-level modeling and verification. To do so, we use a refinement relation *R* defined in [27] that can help to better trace latencies between the different modeling levels.

More precisely, the refinement relation allows us to split tasks into sub-tasks and to replace high-level complexity operators by a concrete sub-behavior. When a task is split into sub-tasks, internal communication must be added between sub-tasks. Figure 5 shows the relation between Partitioning and Software Design models. Latencies are determined between a pair of operators on the same execution path. By performing simulations on either of the two levels, we obtain the occurrences of channel operators (right hand side of the figure) and
can then derive minimum, maximum and average latencies by comparing the times they were reached. The selected latency checkpoints are marked by blue flags on the left hand side of Figure 5, the same in the state machines of TTool diagrams.

4. Contributions

This section discusses how we have further extended our approach from [1, 2] to obtain a more realistic virtual prototype at deployment level. Thus, in order to better capture the behavior of parallel streaming applications, two kinds of extensions are proposed. First, the effects of complex interconnect architectures and resulting high variability of memory accesses should be captured in the model in order to conduct formal verification. Second, the effects of I/O devices that are typical for streaming applications should be captured in sufficient detail to show the impact on performance and allow design space exploration.

4.1. Latency Feedback between Modeling Levels

Latency requirements on channels are annotated by the designer in the TTool diagrams. In our experiments, we focus on average latencies between events. This is due to the fact that latencies vary enormously, depending if memory is accessed in the local or on a distant cluster.

Figure 5: Relation between latencies in Partitioning and Software Design Models from [27]
There are two kinds of problems that are now automatically detected: if the simulation of the virtual prototype at the current abstraction level does not meet the requirement, or if it deviates from more than an indicated percentage, diagrams are annotated accordingly. Deviating latencies appear in red. Model adaptations to solve the problems are however still manual.

4.2. Modeling I/O Co-processors

Input/Output co-processors are hardware or software tasks. In both cases, they take as input a stream of packets, images or video and transform it into a format that the platform can efficiently handle. TTool, up to now, did not provide support for any kind of streaming I/O simulation. We have integrated into TTool high-level representations of some existing models of packet and video streaming co-processors from the same previous work.

The corresponding cycle-accurate SystemC hardware components have now been added to our virtual prototyping environment. In the Deployment Diagrams in [2], we used the existing MWMRCoProc blocks to indicate an interface to the I/O block. The block as such was visible in the block diagram, and was software.

This diagram is no longer semantically correct when a block is hardware. It should then appear on the partitioning level. In the Deployment Diagrams on Software design level, the corresponding blocks now have a new semantics. A corresponding block, called Hardware Accelerator (HWA), becomes available. This somewhat relaxes a paradigm of TTool which requires that no hardware is visible in the Software Design (AVATAR) diagrams, but is necessary to generate the channels leading to and from the interface.

Figure 6: Co-processor interface

The hardware wrapper called VCI MWMR Controller shown in Figure 6 was first presented in [26]. It is used to connect HWA featuring a FIFO in-
interface to the rest of the platform. It provides a VCI initiator and target VCI interface, read and write FIFO interfaces of configurable number, a status and a configuration register.

Wrapper and I/O co-processor instances, signals and netlist as well as the ldscript have to be generated automatically by TTool if they appear in the Deployment Diagram. The fact of adding such detailed models of hardware I/O to the virtual platform thus has significant impact on the generation of top-cell and ldscript and contributes further to the added complexity described in 3.2.

Our contributions, previous and new ones, are now shown throughout a comprehensive case study.

5. Case Study

A particularly interesting telecommunication application is the parallel classification of communication packets, as described in [28]. Task farm properties were captured by Multi-Writer Multi-Reader (MWMR) within channel mapped to shared memory, where an arbitrary number of —hardware or software— tasks access to the same channel.

One of the results of the above work was a detailed analysis of the difficulties and sources of performance loss for this particular application:

- Complex semantics of Multiple-Writer Multiple-Reader channels
- Complex Input/Output Hardware
- High sensibility to non-uniform memory access leading to varying latencies and thus, indirectly, to underflow and overflow of communication channels
- Cyclic behavior of the task graph through feedback of addresses that must be re-used in order to save memory

The task and communication graph shown in Figure 7 features a parallel task graph, where all tasks of a stage $n$ can read the data output by all tasks of stage $n - 1$. In this application, network packets are first cut into chunks of equal size by the input co-processor. Each packet chunk has a descriptor referencing the address of the next chunk. Overall, a chunk contains a 32-bit address, 11 bits to describe its $TotalSize$, 20 bits $date$ for a time stamp, and a boolean $internal$ indicating if the packet is stored on-chip or off-chip, for a total of 64 bits. Only these descriptors are sent through the channels: indeed, packet data are kept in on-chip or off-chip memories. Note that the I/O co-processors have both, VCI and FIFO, interfaces. They also have an Ethernet interface.

The case study tasks are:

- A bootstrap task organizes the system start-up and fills the address channel with a set of addresses generated from the addresses available in packet memory.
Figure 7: Classification application: tasks and communication graph

- An input task constantly reads the addresses initially generated by the bootstrap task. Also, addresses freed from packets having left the system can be reused. Messages between bootstrap, input and output tasks correspond to 32-bit addresses. The Input Engine reads Ethernet encapsulated IP packets, cuts them into slots of equal size, and copies these slots to dedicated memory regions.

- A classification task reads one or several descriptors at a time, then retrieves the first chunk of the corresponding packet from memory. Any classification task can access any chunk. The helper classifiers determine the priority of each packet.

- The scheduling task reads one of the queues according to their priority order, and then writes the descriptor to the output queue. The helper schedulers schedule the packet based on its priority. Both classification and scheduling tasks use try-read primitives to start work whenever data is available and thus maximize performance.

- The output task constantly reads the output queue. Each time a slot is read, the output task frees corresponding addresses, and sends them to the address channel for reuse.

Overall, the application task graph is thus inherently cyclic and very vulnerable to buffer overflows due to contentions.

Thanks to the the latest TTool additions, we are now able to model a much more realistic version of the classification application, including bootstrap and reuse of addresses, non deterministic behavior, I/O operations and monitor of the fill state of channels. Abstraction is made from the actual classification
and scheduling algorithms, where they are replaced by minimum and maximum computation time.

5.1. Modeling at Partitioning Level

Figure 8 shows the design of the application at partitioning level, including the behavior of the Output Engine task. At this stage, the model focuses on the performance properties of tasks and communications. Thus, detailed algorithms and data values are abstracted to computation and transmission complexities. These abstractions facilitate the selection of a HW/SW architecture capable of executing the application.

Communications are modeled as events (in purple) or channels (in blue). Channels are used to send an amount of data, where events are used for synchronization. For example, the Bootstrap and Output Engine first signal to the Input Engine that they will send a new address, and wait for the Input Engine to accept before sending the address.

The data communications are modeled only in terms of their size, but not their values. We model, for example, the size of the packet sent and the size of the packets. When mapped to an architecture, the size of the data communications determines if the current mapping can send the data quickly enough, or if the communications take too long to deliver, and therefore more efficient communication buses should be used.

The abstract behavior of the Output Engine is modeled in terms of its communications and processing complexities. As shown, the Output Engine continually loops. In each loop, the Output Engine waits a certain duration before accepting a packet from the Scheduling. Once it receives a packet, it frees the corresponding address, modeled only as the processing duration at this phase. Once an address is freed, the Output Engine transmits this address to the Input Engine. The data sent and received are modeled only as a number of samples, as our abstraction level does not consider concrete values. Again, the algorithm details and exact data values are expected to be provided only in the Software Design phase, because they are not necessary to take the HW/SW partitioning decisions.

5.2. Modeling at Software Design Level

The circular nature and non-determinism of the application is also reflected in the software design model. Yet, at this level, data types become explicit and task behavior is described at a lower abstraction level.

5.2.1. Application Model

Figure 9 shows the block diagram of the telecommunication application. This architecture shows refined elements such as values exchanged via channels and extra attributes. The packet descriptor is defined as a data type called PacketDesc. The application model features three classification tasks and two scheduling tasks. This relatively low number of tasks is due to current limitations of the graphical representation – AVATAR diagrams cannot yet express
replication of identical tasks and all diagrams and automata have to be designed manually. It still permits realistic modelling of the NUMA and I/O effects; these effects are acerbated with an increasing number of tasks (up to eighty classification tasks were employed, and the channels between the Input Engine and the classification tasks had to be replicated [28]).

All logical channels between blocks are defined as asynchronous. The main communication channel is defined between Classification and Scheduling. It conveys three AVATAR signals that correspond to the three priority queues. Each priority queue (high, medium, low) is translated into a separate multi-writer multi-reader channel in the SoCLib platform.

The upper right part of the window shows three priority queues, one for low, medium and high priority. The priority queues are modeled as asynchronous channels; their depth can be chosen, which here is set to 1024. The width is the item size (here: 8 bytes for a packet descriptor). In the example, channels are configured to be “blocking write” channels.

Figure 10 shows the behavior of the Input and Output Engines modeled with SysML state machine diagrams. The Input Engine first accepts addresses from the bootstrap task. The latter stops working once a preset number of addresses has been reached, at which point the Input Engine accepts addresses fed back from the Output Engine. As shown, the more detailed behavior of the Output Engine is modeled in this phase, including how addresses are freed. The after represents an estimation of the duration of the scheduling algorithm after model refinement.

The model of the classification task is rather complex. It contains three identical classification sub-tasks, which get work from the from_IE channel in a task-farm manner. This task-farm way is modeled by non-deterministic choice modeled in the state Waiting. Each classification sub-task then classifies the packet, and a guarded choice operator selects the low, minimum or high priority queue. Figure 12 shows one of the inner classification tasks. Priority queues
behavior are modeled inside an outer classification task. Figure 11 shows the state machine diagram of the outer classification task. This model is very complex because we have to reflect that either we read a packet from Input Engine, dispatch it to one of the classifiers and go back to the WaitForPacket state (left hand side), or read packet already classified from one of the inner tasks and then transmit it (center) or there still may be available packets in the channels filled up by the inner tasks (right hand side).

The scheduling tasks which take up work are modeled in a similar way (Figures 13 and 14). An outer task coordinates the reading from the priority queues and the writing to the output channel. The scheduling tasks are somewhat less complex because they write to a common output queue and need not contain choice operators.

5.2.2. Mapping

Allocation of tasks and channels onto the target MPSoC is explicitly captured within deployment diagrams (Figure 15). These diagrams contain hardware nodes (e.g., CPUs and memory banks) that can be customized with parameters such as cache associativity, memory size, network latency, etc. Software tasks are mapped onto the execution nodes of the platform, and channels between tasks are mapped onto the memories.

I/O can be modeled in two ways: either as software tasks or as SoC Lib models of hardware accelerators. Packet processor models are taken from [16]: one is used for reading packets from either an Ethernet connection or a file,
and the other is used for writing packets after computations. The more recent network I/O facilities of SoCLib are used, allowing the reading in and analysis of Ethernet encapsulated IPv4 packets.

Timers and interrupts are not modeled in the deployment diagram because they are automatically added to the prototyping model during the generation of the SoCLib virtual platform. There are six processors in total. The target architecture features a two-level interconnect based on a Virtual Generic Micro Network (VGMN), which behaves as two independent packet switched networks for commands and responses, and two local crossbars.

6. Experiments

Parallel packet classification is known to suffer from several major performance impediments, such as overflow of the input channels for high throughput, cache misses and conflicts when taking locks, resulting in high latencies. We explore different mappings of the classification application featuring three classifiers, three priority queues, and two schedulers. Compared to our previous contributions, our experimentation is no more limited to the software part of the application.

6.1. Test-bed

As we now model the (nearly) complete application, the generated platform should yield results comparable to the real one with 3 classifiers, 2 schedulers, no bursts and no packet memory access.

In the experiments, we focus on measuring latencies to analyze the performance of our system. After the Input Engine finishes reading addresses from the Bootstrap task, it reads addresses which have been liberated by the Output Engine. By inserting checkpoints in the channel leaving the Input Engine and the one entering the Output Engine, we can thus measure the processing time of such fed back addresses. Figure 10 shows the state machine diagrams annotated with latency checkpoints in the form of small flags.

Figure 10: State machines of the Input Engine and the Output Engine with latency checkpoints
6.2. Assumptions

There still are some restrictions to our model. For example, we do not model packet storage in these experiments, neither do we represent burst transfer of packet descriptors to obtain better performances. Since we are interested in general packet timing, the algorithms inside the tasks are still quite abstracted with regards to the real ones (which serve as time estimates), as we are more interested in timings rather than detailed application behavior.

6.3. Partitioning Level

Latency checkpoints are added to the Partitioning models: one is placed in the Input Engine (as soon as the bootstrap is finished and the application runs in steady state) and the corresponding one in the Output Engine (as soon as the packet is read, the after annotation reflects an estimation of the duration of the computation), as shown previously in Figure 8. This enables us to compare high-level simulation results with results obtained by cycle-bit accurate simulations.

6.4. Software Design Level

Channels are stored in memory, and the communications via these channels represent a high fraction of the application activity. When the number of tasks
accessing a channel increases, the amount of time spent waiting for the lock also increases.

The experiment is based on the cycle/bit accurate level of SoCLib, thus with a high level of precision, at the price of rather slow simulations. General purpose processors simulated in the virtual prototype are PowerPC 405 running at 433MHZ. The processors run the MutekH [29] micro kernel. For reading of actual Ethernet packets, we rely on the network_io API provided with SoCLib. The abstract models of Input and Output Engine assume an equal part of high, medium and low priority packets, modeled by a Random operator.

In the virtual prototype, spy points —represented by magnifying glasses—correspond to hardware probes which log all transfers on the VCI interconnect. Figure 15 shows how this spy mechanism, described in [30], is now integrated in the deployment diagram. The top cell automatically generated from deployment diagrams integrates the corresponding logger/statistics modules. The input channel (InputEngine/out packet, spyglass 1), is mapped on the I/O cluster and accessed by the classification tasks from cluster 0. The three priority queues (Classification/out queue_low, spyglass 2), mapped on Memory0, are written by
three classification tasks and read by two scheduling tasks. The output channel `Scheduling/out packet`, is mapped onto the scheduling cluster and monitored by spy 3, the address feedback channel (`OutputEngine/out address`) by spy 4.

### 6.5. Experimental Results

We show results for three different mappings. All take into account the results of [30] that the `Scheduling/out packet` channel must be mapped on the same cluster as the scheduling tasks.

1. **Mapping 1**: In the mapping presented in Figure 15, three clusters are used, one for the classification tasks, one for the scheduling tasks, the final one for I/O.
2. **Mapping 2**: As shown in Figure 16, I/O tasks are mapped on the classification cluster. The channel containing the packet descriptors produced
by the Input Engine is mapped on RAM2 in the I/O cluster, whereas the channel containing the liberated addresses is mapped on RAM1 of the scheduling cluster. This choice was made because there is plenty of time left for an address to return to the input engine. Addresses are small, so buffers can thus be dimensioned with a high depth, and the latency to cross over from one cluster to another is compensated.

3. **Mapping 3**: The last mapping features the same mapping as in Figure 16, except that the classification/bootstrap cluster contains an additional memory bank in order to store the address buffers (not shown).

Table 1 compares the results between the partitioning (high-level transactional simulation) and software design levels. On the software design level, we show results for the two variants: one where the very abstract models of Input and Output Engine are used, from which software tasks are generated, and one with the actual SoCLib models (Input Engine consisting of more than 600, the slightly simpler Output Engine of more than 400 lines of SystemC code). The series of measurements show latencies (in milliseconds) from the reception of a new or recycled address by the Input Engine to the reception of the computed packet by the Output Engine (`receive(address) -> receive(packet)`), for the three mappings. These are averages for all paths taken between the two operators in question.

We can observe that standard deviations are significant for the “end-to-end” latencies, as packet descriptors might or might not be blocked due to competition for access to a memory bank containing a channel.

Our first parameterizing of the hardware on the partitioning level actually led to an overestimation of latencies by a factor of 6 to 8 (not shown). We therefore modified our application model by reducing delay estimations to take into account the much lower latencies obtained on the virtual prototype. The columns **Partitioning** show latencies after these corrections. For the latencies measured in **Software Design**, however, the second mapping demonstrates the worst performance, which is not observed in the Partitioning evaluations.

The rightmost three columns of Table 1 show increasingly realistic virtual prototypes, the rightmost obtained with the original one from [30], featuring
handwritten C Posix code for these tasks. When replacing the Input and Output Engine generated from the abstract description by the cycle-bit accurate (caba) models, we note a slight increase of latencies, thus would have revise the abstract models shown in Figure 10 and add delays in order to be closer to the more realistic model. We finally compare our prototypes, where the code of the classification and scheduling tasks is generated from timed automata, handwritten C Posix code for these tasks, for the same configuration and number of tasks and channels (rightmost column).

It should be noted that we do not model packet memory, thus have fewer source of contention when accessing the memory bank, and thus a slight advantage wrt. the original application. The results obtained with SoCLib models of the I/O co-processors are clearly closer to the results for the complete system with handwritten Posix tasks.

However, while a simulation on partitioning level is performed in a matter of seconds, the full-system simulation of the platform takes several hours (execution of the bootstrap and transmitting 10k packets of random priorities) even with a relatively small number of tasks. As shown in [2] where we focused on throughput, in the 20 million simulation cycles after bootstrap $3.5 \times 10^{-5}$ bytes are transferred per simulation cycle. With the cycle-bit accurate models of the I/O engines, simulation time remains practically unchanged: the number of transfers on the VCI interconnect dominates the time for cycle accurate simulation of I/O engines. Transfers initiated by software running on CPU are replaced by the same transfers initiated by HW co-processors.

<table>
<thead>
<tr>
<th>Mapping</th>
<th>Partitioning</th>
<th>Virtual Prototype</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>avg</td>
<td>std dev</td>
</tr>
<tr>
<td>Mapping 1</td>
<td>18.6</td>
<td>12.2</td>
</tr>
<tr>
<td>Mapping 2</td>
<td>19.2</td>
<td>11.1</td>
</tr>
<tr>
<td>Mapping 3</td>
<td>17.9</td>
<td>11.2</td>
</tr>
</tbody>
</table>

### 6.6. Latency Feedback on diagrams

As was mentioned in section 4.1, latencies are annotated and eventual deviations reported. Figure 17 shows an example of latency annotations on the software design level (the signal is internally numbered 752). In the example, we measured an average latency of 327 microseconds from the reception of a address by the Input Engine to the reception of this address by the Output Engine.

### 6.7. Discussion

Due to current restrictions on the number of tasks which can be represented in TTool diagrams, the number of tasks is small, but still sufficient to demonstrate the effects of contention. In previous work on the same application, the
effects shown here were exacerbated as experiments were performed for dozens of classification tasks, mapped to several different clusters and thus creating even higher contention on the interconnect.

An important advantage of working with a graphical interface like TTool is that the rather important structural modifications between the mappings (passing from three to two clusters and redistributing channels) took less than five minutes. In the original exploration, shell scripts could impact only the numbers of tasks and memory banks, so such structural changes of the underlying hardware would require a rewriting of the top-cell.

What we are still lacking is the determination of latencies for individual packets. In order to obtain this, an identifier has to be inserted into the descriptor by the Input Engine and extracted by the Output Engine. We could also gain more precise results on the Software design level by inserting actual C Posix code into the description by timed automata—for this purpose, AVATAR features a prototyping option—but this would mean that generated application code is no longer correct by construction.

7. Conclusion and Future Work

This paper extends a UML/SysML virtual prototyping environment with new modeling and evaluation capabilities for streaming applications. We first explain the necessary extensions: modeling extensions, semantics extension, and code generation extensions. We then show how these extensions have been integrated into TTool. We also demonstrate how performance metrics such as latency can be estimated, then measured, on models spanning different levels of abstraction.

An exhaustive case study shows the modeling and experimental results for an application that is particularly difficult to handle by regular tools: high performance packet classification. We obtain a generated platform and application very close to the original, handwritten one by extending TTool to handle the new aspects: task-farm, NUMA, I/O.
High performance streaming applications such as the one analyzed in the case study usually feature a larger number of tasks; we will thus require the ability to describe and automatically generate multiple identical tasks, and to generate code for these tasks accordingly. As shown for the example featuring three classification and two scheduling tasks, the complexity for obtaining a semantically correct model is high and, for larger numbers of tasks, near impossible to model by hand.

Other typical I/O co-processors should be made available in the deployment diagrams of TTool. The problem is that such components are very specific and corresponding SoCLib models do not always exist. The generation of virtual co-processors on different degrees of abstraction, starting at partitioning level, thus is another interesting and important issue.

Our long term work targets full design space exploration. Even if currently re-mapping is only semi-automatic and requires the designer to manually adapt the architecture and the mapping, the level of abstraction allows for changes to be made quickly and easily. More importantly, profound structural changes are modeled and reflected in the virtual prototype in a matter of minutes.

References


