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Abstract—Desirable application performance is typically guaranteed through the use of Service Level Agreements (SLAs) that specify fixed fractions of resource capacities that must be allocated for unencumbered use by the application. The mapping between what constitutes desirable performance and SLAs is not unique: multiple SLA expressions might be functionally equivalent. Having the flexibility to transform SLAs from one form to another in a manner that is provably safe would enable hosting solutions to achieve significant efficiencies. This paper demonstrates the promise of such an approach by proposing a type-theoretic framework for the representation and safe transformation of SLAs. Based on that framework, the paper describes a methodical approach for the inference of efficient and safe mappings of periodic, real-time tasks to the physical and virtual hosts that constitute a hierarchical scheduler. Extensive experimental results support the conclusion that the flexibility afforded by safe SLA transformations has the potential to yield significant savings.

I. INTRODUCTION

Motivation and Scope: Formal verification of the safety properties of software systems has long been the “holy grail” of a number of research communities focusing on safety-critical applications. While significant progress has been achieved in highly-specialized domains (such as for automotive [1] and avionics [2] applications), making formal verification accessible to system builders remains elusive. In this paper, we address two issues related to this state-of-affairs. First, we show that the use of “lightweight” formalisms to support the reasoning processes of system builders (programmers, integrators, and operators) can be quite effective as an alternative to approaches that require significant, deep knowledge of the specific application domain and/or those that require fluency in a particular formalism. Second, we show that the notion of “safety” extends well beyond systems in which loss of human life or expensive equipment is at stake. In particular, we consider a cloud setting wherein it is desirable to safely, yet efficiently aggregate (co-locate) workloads that are subject to contractual Service Level Agreements (SLAs). In this setting, the system/cloud operator cannot be expected to know (let alone be an expert in) the purpose from a particular SLA request, and thus cannot be expected to arbitrarily modify the set of SLAs for efficient co-location purposes. Rather, the only degree of freedom that such an operator could be assumed to have is the ability to transform an SLA from one form to another, as long as the transformation can be formally verified to be equivalent with respect to a given SLA calculus – a calculus which is independent from and does not require expertise in the application domains of the constituent workloads.

Safe SLA Transformations for Efficient Co-location: The wide proliferation and adoption of virtualization technologies can be attributed to the various benefits they deliver, including cost efficiency (through judicious resource consolidation), deployment flexibility (through just-in-time “cloud” resource acquisition), simplified management (through streamlined business processes), among others. Virtualization delivers these benefits thanks in large part to a key attribute: performance isolation – the ability of a user (or a set of applications thereof) to acquire appropriate fractions of shared fixed-capacity resources for unencumbered use subject to well-defined, binding service-level agreements (SLAs) that ensure the satisfaction of minimal quality of service (QoS) requirements. In many instances, however, multiple SLA forms may be “equivalent” in terms of their ability to support a given QoS. For instance, a QoS that spells out an upper bound on (say) data availability in a storage solution could be expressed by any number of SLA forms, depending on how device reliability and redundancy are combined. Similarly, a QoS that spells out an upper bound on (say) missed deadlines in a real-time system could be satisfied by any number of SLA forms, depending on reservation granularity and/or on underlying schedulers. The ability to transform (or rewrite) SLAs from one form to another safely (i.e., without jeopardizing the safety or QoS requirements of an underlying workload) gives cloud operators significant freedom in efficiently managing their resources.

A Framework for Safe SLA Transformation: We present our ideas for formal verification of SLA transformations within the context of the generic framework depicted in Figure 1. Our framework enables a user (e.g., a cloud service operator) to propose SLA transformation rules, which upon successful verification for soundness by the AARTIFACT system [3] are fed to an SLA transformation engine. The SLA transformation engine utilizes a set of rewriting rules from the “Transformation Rules Repository” to explore the space of possible safe co-location configurations (mappings). If a feasible mapping is found,\(^1\) then the transformation engine informs the

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\(^1\)Naturally, if multiple feasible mappings are found, then it is possible to use a scoring function to select the optimal mapping to realize.
While our safe SLA transformation framework depicted in Figure 1 is quite general, ultimately, any instantiation of this framework must be based on a specific “calculus” that enables the manipulation of SLAs. By “calculus”, we mean a model that enables reasoning for the purpose of establishing SLA relationships/equivalencies. In this paper, we use a specific calculus that is particularly suited for periodic real-time systems [5]. In particular, our calculus is defined for a setting in which the “Task Workload” consists of a set of periodic real-time tasks, and the set of “Available Resources” are fixed-capacity hosts, which are individually scheduled using a Rate Monotonic Scheduler (RMS) [6]. Here we note that Our RMS-based calculus can be viewed as a simplified variant of the real-time calculus [7]. In general, our framework admits the use of other calculi (e.g., linear algebra or network calculus) if an appropriate database of propositions is supplied to the AARTIFACT system.

Paper Outline: The remainder of this paper is organized as follows. In Section II, we overview the various building blocks of our framework. In Section III, we present the particular domain-specific language that we adopt in this paper for the expression of SLAs for real-time workloads. In Section IV, we present examples of SLA transformations with associated machine-readable proofs of correctness that were mechanically verified using the AARTIFACT system. We conclude in Section V with a discussion of related work.

II. OVERVIEW OF FRAMEWORK BUILDING BLOCKS

In this section, we present the basic background necessary to understand the underpinnings of the SLA transformation engine and the resource manager elements of the framework depicted in Figure 1. We follow that with an overview of the salient features and capabilities of the AARTIFACT assistant used in our framework.

A. Real-Time Resource Management

The problem of efficiently co-locating tasks on a set of hosts is effectively a “packing” process – given a set of tasks (the workload), a scheduler packs these tasks into groups, such that the SLAs of all tasks in a given group can be satisfied by a single host. For non-real-time tasks, a set of tasks can be packed in a single group as long as the total utilization needs of all tasks in a group is less than the capacity of the host. For real-time tasks, whether or not a set of such tasks can be packed in a single host depends on other details – including the scheduler deployed in the host – and is generally more involved.

Liu and Layland [6] provided classical results for the schedulability condition of \( n \) periodic real-time tasks, each requesting a resource for \( C_i \) units of time every \( T_i \) units of time (the period). Specifically, assuming a preemptive priority scheduler and a rate-monotonic assignment of priorities to tasks, a group of \( n \) tasks is schedulable (i.e., could be packed) on a single host if \( \sum_{i=1}^{n} \frac{C_i}{T_i} \leq n(\sqrt{2} - 1) \). Follow-up work [8], [9] showed that if the set of \( n \) tasks can be grouped into \( k \) subsets such that the periods of tasks in each subset are multiples of one another (i.e., harmonic), then a tighter schedulability condition is \( \sum_{i=1}^{n} \frac{C_i}{T_i} \leq k(\sqrt{2} - 1) \).

A schedulability condition provides us with the means to check whether a specific set of tasks is schedulable on a single host. In our setting, however, the problem is to partition a large set of tasks (the workload) into smaller sets, each of which would be schedulable on a single host. Moreover, in our setting, safe SLA transformations imply that the specification of the resource requirements (e.g., \( C_i \) and \( T_i \)) for each task may not be unique. Rather, the SLA for each task could be satisfied by multiple resource requirement specifications. An optimal partitioning of the workload (namely, one that uses the least number of hosts) can be shown to be NP-hard, in

\(^2\)Without loss of generality, in this paper, we assume that all hosts are identical.
B. Underlying Verification System and Proposition Database

AARTIFACT is a formal reasoning assistant\(^4\) that supports lightweight verification of formal arguments using a database of concepts, propositions, and syntactic idioms that deal with common mathematical concepts such as numbers, vectors, and sets. It has a familiar concrete syntax overlapping with English, MediaWiki markup, and \LaTeX, and a friendly user interface [3]. The AARTIFACT system has been used in the past to formally verify the consistency [12] of a formal framework of safe transformations of constraints governing constrained-flow networks [13].

The system’s flexible design allows domain expert managers to quickly and easily assemble a large database for a specific domain. They can do so by employing custom natural language constructs\(^3\) in combination with formal syntax, such as in the example below.

\[
\begin{align*}
\text{for all } a, T, T', C, \\
T' & \geq T \\
\text{and for all } a' \text{ length } T \text{ subvector of } a, \ (\sum a') & \geq C \\
\text{implies} & \\
\text{for all } a' \text{ length } T' \text{ subvector of } a, \ (\sum a') & \geq C.
\end{align*}
\]

The following example illustrates the kind of facts about common mathematical concepts that can also be included.

\[
\begin{align*}
\text{for any } x, y, z, \\
x & \in \mathbb{R}, \ y \in \mathbb{R}, \ z \in \mathbb{R}, \ x < y, \ y < z \\
\text{implies} & \\
x & < z.
\end{align*}
\]

It is also possible to encode an equivalence between two forms of notation or syntax. They can be viewed as establishing a normal form for representing certain concepts or properties thereof. For example, the following proposition converts the typical notation for a sum of a finite range of components in a vector, “\(v_i + \ldots + v_j\)”, into a predicate that is then used in other propositions about the properties of this concept.

\[
\begin{align*}
\text{for any } \ v, i, j, \\
\ v & \text{ is a vector, } \ v \in \mathbb{Z}^{|v|}, \ 0 \leq i, \ i \leq j, \ j < |v|, \\
\text{implies that} & \\
v_i + \ldots + v_j & \text{ is the sum of components in} \\
\ v & \text{ in the index range } i \ \text{to} \ j.
\end{align*}
\]

AARTIFACT’s expert-managed database of propositions and syntactic idioms is compiled into two separate applications. The first application is a server-side verifier that can be executed when requests are submitted to it. It parses arguments authored using a familiar concrete syntax that consists of common \LaTeX macros and English phrases [14]. It validates these arguments using an inference algorithm the context of which has been augmented with a data structure that uses the database of propositions to compute congruence closures [15] throughout the verification process. The AARTIFACT system’s inference algorithm relies on a collection of inference rules corresponding to those found in a typical definition of higher-order logic [16] (i.e. those governing conjunction, disjunction, negation, and quantification), and variants thereof found in common sequent calculus formulations [17]. The validation procedure verifies arguments with respect to these inference rules as well as the entire database of propositions. The verification capabilities provided by the AARTIFACT system are “lightweight” in that it is not essential that any guarantee be provided about the logical consistency or completeness of the validation process (allowing a user to validate that, for example, the proper syntax is used, that all variables are bound, or that only “intuitive” symbolic manipulations are employed). However, the soundness of the validation process can be guaranteed if the system can use only a subset of the inference rules and database propositions that is consistent with a particular logic. This capability has been demonstrated for propositional and first-order logic, and the proofs verified in this paper are verified using a consistent subset of the propositions that includes many facts about the properties of vectors and sets.

The second application is a client-side JavaScript that automatically informs users of syntactic idioms available in the ontology as they author formal arguments [18]. This capability is essential as it informs users of which custom natural language constructs have been introduced by expert managers and are available for use in formal arguments.

III. A DOMAIN-SPECIFIC LANGUAGE FOR REAL-TIME SLAS

As we hinted earlier, SLA transformations must be based on a given “calculus". Thus, for the purposes of SLAs governing real-time workloads (and without loss of generality), in this section we present a specific “language" for expressing SLAs and transformations thereof. We note that it is the responsibility of the domain expert (cf. Figure 1) to define and codify the SLA “language" as part of the AARTIFACT system’s database of propositions. In section IV, we provide examples which

\(^3\)Our problem can be easily reduced to a multiprocessor real-time scheduling problem, which is known to be NP-hard [10].

\(^4\)An interactive demonstration of the system is available online at http://www.aartifact.org.

\(^5\)For readability, we underline custom natural language constructs utilized within formal expressions.
A schedule is an infinite sequence (or vector) of 0s and 1s in which each entry represents whether or not a resource is allocated to the single task in question at that discrete point in time (that is, the allocation deadline is T units of time from the beginning of the period).

As presented, our SLA types are general enough to express a wide range of supply/demand elements. For example, an SLA of type \((C, T, D, W) \in \mathbb{N}^4, C \leq T, D \leq W, \text{ and } W \geq 1\), where \(C\) denotes the resource capacity supplied or demanded in each interval of time that is \(T\) units in length, and \(D\) is the maximum number of times such an allocation cannot be honored in a window consisting of \(W\) allocation intervals.

Definition. A hard SLA type \(\tau\) is a tuple of natural numbers \((C, T, D, W) \in \mathbb{N}^4, C \leq T, D \leq W, \text{ and } W \geq 1\), where \(C\) denotes the resource capacity supplied or demanded in each interval of time that is \(T\) units long; such a type requires that the allocation of resources is honored at all times.

Note that these definitions imply that a type \((C, T)\) is equivalent to \((C, T, 0, 1)\). These definitions assume that the periodic capacity could be allocated as early as the beginning of any interval (or period) and must be completely produced or consumed by the end of that same interval (that is, the allocation deadline is \(T\) units of time from the beginning of the period).

As presented, our SLA types are general enough to express a wide range of supply/demand elements. For example, an SLA of type \((1, 1, 0, 1)\) could be used to characterize a uniform, unit-capacity supply provided by a physical host. An SLA of type \((1, n, 0, 1)\), \(n > 1\) could be used to characterize the fractional supply provided under a general processor sharing (GPS) model to \(n\) processes. An SLA of type \((1, 30)\) could be used to represent a task that needs a unit capacity \(C = 1\) over an allocation period \(T = 30\) and cannot tolerate any missed allocations. An SLA of type \((k, k \cdot n, 0, 1)\), \(n > 1, k \geq 1\) can characterize the fractional supply provided in a round robin fashion to \(n\) processes using a quantum \(k\). In all of these examples, the SLA type does not admit missed allocations (since \(D = 0\)). An SLA of type \((1, 30, 2, 5)\) is an example of a task that tolerates missed allocations (soft deadline semantics) as long as there are no more than \(D = 2\) such misses in any window of \(W = 5\) consecutive allocation periods.

B. Strong Satisfaction of Hard SLAs

We consider what it means to satisfy an SLA type. The following definitions formalize the notion of satisfaction by considering a schedule that governs a single task. We first do so for SLAs of the form \((C, T)\) (i.e., those that do not admit missed allocations).

Definition. A schedule \(a\) is a vector \(a \in \{0, 1\}^\infty\).

A schedule is an infinite sequence (or vector) of 0s and 1s in which each entry represents whether or not a resource is allocated to the single task in question at that discrete point in time. We use the subscript notation \(a_i\) to refer to an entry in the sequence corresponding to a particular point in time. A schedule \(a\) is said to strongly satisfy (denoted by \(\models_s\)) a hard SLA type \((C, T)\) if the resource is allocated for \(C\) units of time in all (overlapping) intervals of length \(T\).

Definition. We say \(a \models_s (C, T)\) iff for every \(m \geq 0\),

\[a_m + \cdots + a_{m+(T-1)} \geq C\]

C. Weak Satisfaction of Soft SLAs

We generalize the above definition for the more general soft SLA types of the form \((C, T, D, W)\). To do so, we introduce an aggregate vector \(A(a, C, T)\) that characterizes a schedule \(a\) with respect to some \(C\) and \(T\). We also switch over to a notion of weak SLA satisfaction.\(^6\)

Definition. \(A(a, C, T) \in \{0, 1\}^\infty\) is a vector defined as:

\[A(a, C, T)_m = \begin{cases} 0 & \text{if } a_m + \cdots + a_{m+(T-1)} < C \smallskip \text{1} & \text{if } a_m + \cdots + a_{m+(T-1)} \geq C \end{cases}\]

A schedule \(a\) is said to weakly satisfy (denoted by \(\models_w\)) an SLA type \((C, T, D, W)\) if the resource is allocated for \(C\) units

\(^6\)Our SLA model mirrors existing periodic task models in the real-time scheduling literature [19]–[21].
of time in at least $W - D$ out of every $W$ non-overlapping (contiguous) intervals of length $T$.

**Definition.** $a \models_w (C, T, D, W)$ iff for every $Q \in \mathbb{N}$ and $m = Q \cdot (W \cdot T)$,

$$W - D \leq \sum_{j=0}^{W-1} A(a, C, T)_{m+Tj}.$$  

**IV. Machine-assisted Verification of SLA Transformations**

In this section, we present a set of SLA transformations that exemplify (and certainly do not exhaust) the range of conjectures that a user of our framework (cf. Figure 1) would need to formally check for safety (before allowing the SLA transformation engine to use these transformations). Recall that the verification of these transformations must be based on specific assumptions regarding the underlying scheduler. As we mentioned before, in this paper we assume that the underlying scheduler is a Rate Monotonic Scheduler.

By assembling these proofs, the system user can be certain that her assertions about the safety of SLA transformations are grounded in the semantics of integer vectors (encoded in the AARTIFACT proposition database) that underlies the definitions of SLAs.

**A. Example: Strong Satisfaction of Hard SLAs**

Each SLA transformation is a type inference rule associated with a set of pre-conditions. Intuitively, each inference rule implies that we can safely substitute (rewrite) some SLA type $(C, T)$ for some other type $(C', T')$ as long as the pre-conditions of the inference rule are met. We present the inference rules for rewriting hard SLAs in Figure 2.

In order to establish the soundness of these inference rules for a particular satisfaction relation (i.e., $|=s$ or $|=w$), we must provide a proof for each rule. In this context, a rule is *sound* if it is consistent with the underlying AARTIFACT database of propositions (governing vectors, sets, and integers) that has been assembled for this application domain. We first present a simple mechanically verifiable proof of the soundness of one of the rewriting rules in Figure 2 with respect to strong satisfiability. All mechanically verifiable (using AARTIFACT) arguments are enclosed in boxes.

We first reproduce below the machine-readable definitions for hard SLA types and strong satisfiability.

**Assume for any $C, T \in \mathbb{N}$,**  

$(C, T)$ is a hard SLA type iff $C \leq T$.

**Assume for any $a, C, T$,**  

$a$ strongly satisfies $(C, T)$ iff  

- $a \in \{0, 1\}^\infty$,  
- $(C, T)$ is hard SLA type,  
- and for all $a'$ length $T$ subvector of $a$, $(\sum a') \geq C$.

This definition associates the notion of an SLA with a proposition about integer vector. The definition quantifies over the subvectors of a vector using the custom natural language construct “length $n$ subvector of $v$” made available by the domain expert. The prefix operator “$\sum$” is used to denote the sum of the elements of a vector of integers.

By assembling these proofs, the user can now assemble a proof that the $\text{[TIME-1]}$ rewriting rule is safe with respect to strong satisfaction.

**Theorem 1.** The $\text{[TIME-1]}$ rule is sound with respect to strong satisfaction.  

**Proof:** Given the database of propositions about vectors assembled by expert managers, the proof is a trivial unwinding of definitions.

**Assume for any $a, C, T$,**  

$(C, T)$ is a hard SLA type and  

$a$ strongly satisfies $(C, T)$  

implies  

for all $T' \in \mathbb{N}$, $T' \geq T$ implies  

$(C, T')$ is a hard SLA type,  

for all $a'$ length $T$ subvector of $a$, $(\sum a') \geq C$,  

for all $a''$ length $T'$ subvector of $a$, $(\sum a'') \geq C$,  

$a$ strongly satisfies $(C, T')$.

**B. Example: Weak Satisfaction of Soft SLAs**

The set of SLA rewrite rules in Figure 2 did not allow transformations that involve the manipulation of SLAs with non-trivial $D$ and $W$ parameters. In Figure 3, we provide another rewriting rule that is applicable to soft SLA types.  

The new, third rule $\text{[DROPS-1]}$ allows for adjustment of the $D$ parameter. Since we are now dealing with weak satisfiability of soft SLAs, we need the corresponding definitions for these in terms of integers vectors.

**Assume for any $C, T, D, W \in \mathbb{N}$,**  

$(C, T, D, W)$ is a soft SLA type iff  

$C \leq T$, $D \leq W$, and $W \geq 1$.

Assume  

for any $C, T, D, W \in \mathbb{N}$,  

for any $a$,  

$a$ weakly satisfies $(C, T, D, W)$ iff  

- $a \in \{0, 1\}^\infty$,  
- $(C, T, D, W)$ is a soft SLA type,  
- for all $a'$ in split $a$ by $T \cdot W$,  
- $|(a'' \text{ in split } a' \text{ by } T) | \leq D$.

---

8Note that these can be used in conjunction with the rules in Figure 3 if the rules in Figure 3 are appropriately interpreted as having $D = 0$ and $W = 1$ and are valid for weak satisfaction.
Notice that the system user employed a custom natural language construct “in split v by n” made available by the domain expert. This construct allows the system user to reason about a common, intuitive operation on vectors: splitting a vector v into contiguous subvectors of length n. The system user can now argue that the [DROPS-1] rule is safe given with respect to weak satisfaction.

**Theorem 2.** The [DROPS-1] rule is sound with respect to weak satisfaction.

**Proof:** Once again, the proof is a matter of unwinding definitions.

Assume for any \( a \in \{0,1\}^*, C, T, D, D', W \in \mathbb{N} \),

\[
(C, T, D, W) \text{ is a soft SLA type, and} \quad a \text{ weakly satisfies } (C, T, D, W)
\]

implies

\[
(C, T, D', W) \text{ is a soft SLA type, and} \quad a \text{ weakly satisfies } (C, T, D', W).
\]

The system user can also build arguments about specific cases. The [D-1-1] rule deals with the specific case in which exactly one out of every two time spans does not provide sufficient resources. The system user constructs a proof that in this case, it is possible to set \( T' = W \) to ensure that \( D = 0 \).

The argument first unwinds the definitions, as before. Next, it is shown that the overall sum of the elements of each subvector of size \( W \) is at least \( C \) if at least one out of every two subvectors of size \( T \) has sum at least \( C \). The verification system is able to recognize this line of reasoning because this fact is among those included by the domain expert in the database of propositions. The user then indicates that splitting \( a' \), a vector of length \( T \cdot 2 \), into subvectors of length \( T \cdot 2 \) yields a set containing only \( a' \). Since the sum of the elements of \( a' \) already exceeds \( C \), all the elements of this split trivially have sum at least \( C \). This means no elements have sum less than \( C \). After a few more substitutions, the definition of weak satisfaction is restated for the new parameters. Note that at several steps in the formal argument, multiple substitutions are applied simultaneously. Notice also that no explicit annotations are provided to the verifier justifying each step. These capabilities of the AARTIFACT system are designed specifically for the convenience of the system user.

**Theorem 3.** The [D-1-1] rule is sound with respect to weak satisfaction.

**Proof:** This proof requires a larger number of reasoning steps.

Assume for any \( a \in \{0,1\}^*, C, T, T' \in \mathbb{N} \),

\( T' = T \cdot 2 \), \( (C, T, 1, 2) \) is a soft SLA type, and

\( a \) weakly satisfies \((C, T, 1, 2)\)

implies

\( (C, T, 0, 1) \) is a soft SLA type, and

for all \( a' \) in split \( a \) by \( T \cdot 2 \),

\[
\{a'' \text{ in split } a' \text{ by } T \mid \sum a'' < C \} \leq 1,
\]

\[
\{a'' \text{ in split } a' \text{ by } T \mid \sum a'' \geq C \} \geq 1,
\]

\[
\sum a'' \geq C,
\]

\[
\{a' \} = T \cdot 2,
\]

\[
\{a'\} = \{a'' \text{ in split } a' \text{ by } T \cdot 2 \},
\]

\[
1 = \{a'' \mid \sum a'' < C \}
\]

\[
T \cdot 2 = T'' \cdot 1,
\]

\[
\{a'' \text{ in split } a' \text{ by } T'' \cdot 1 \mid \sum a'' < C \} \leq 0,
\]

for all \( a' \) in split \( a \) by \( T'' \cdot 1 \),

\[
\{a'' \text{ in split } a' \text{ by } T'' \mid \sum a'' < C \} \leq 0,
\]

\( a \) weakly satisfies \((C, T', 0, 1)\).

**V. Related Work**

The work described in this paper brings to bear the premise of formal verification using the AARTIFACT lightweight proof assistant on the problem of efficient co-location of real-time workloads. We believe that the capabilities and flexibility enabled through this framework are crucial for emerging shared infrastructure cloud and grid environments, and as far as we can tell there is no prior work that proposed or developed such capabilities. That said, there are two distinct bodies of related work to consider – namely those concerned with applying real-time scheduling theory to real-time workload management, and those concerned with the development of
formal verification systems and their practical application within specific domains.

**Efficient Co-location of Real-Time Workloads:** In a cloud or grid setting, the hosting environment operates at a macroscopic scale (e.g., enforcing specific resource utilization ratios over relatively long time scales). While appropriate for many applications, such coarse SLAs do not cater well to the needs of real-time applications, whose QoS constraints require resource allocations at a more granular scale, e.g., through the specification of a worst-case periodic resource utilization. A very effective mechanism for dealing with this mismatch is the use of hierarchical scheduling, whereby the granularity of the reservations is refined as virtualization layers are traversed. Using hierarchical scheduling, resources are allocated by a parent scheduler at one level of the hierarchy to a child scheduler (or a leaf application) at the next level of the hierarchy. Conceptually, at any given layer of this hierarchy, the parent scheduler can be seen as allocating a virtual slice of the host at some granularity which is further refined by lower-layer schedulers, until eventually appropriated and consumed by a leaf application.

Numerous previous studies dealt with hierarchical scheduling frameworks [22]–[25]. Regehr and Stankovic [22] introduced a hierarchical scheduling framework in support of various types of guarantees. They use rewriting rules to convert a guarantee provided under a specific scheduling algorithm to a guarantee provided under another. This notion of rewriting is different from ours as it does not accommodate workload transformations. Shin and Lee [23] present a compositional real-time scheduling framework based on workload bounding functions, and resource bounding functions. They utilize a tree data structure, where a child scheduling system is the immediate descendant of the parent scheduling system, and the parent scheduling system is the immediate ancestor of the child scheduling system. Their model assumes that the parent and children scheduling systems can utilize different types of scheduling algorithms. Under their framework any given system composed of a workload, resources, and a scheduling algorithm will be schedulable if the minimum resource curve bounds the maximum workload curve. This model has also been extended further [24] to include context switching overhead and incremental analysis.

A common assumption in hierarchical scheduling frameworks is that the grouping of applications and/or schedulers under a common ancestor in the scheduling hierarchy is known *a priori* based on domain specific knowledge; for example, all applications with the same priority are grouped into a single cluster, or all applications requiring a particular flavor of scheduling are grouped into a single cluster managed by the desired scheduling scheme. Most of this prior body of work is concerned with addressing the schedulability problem given such a fixed hierarchical structure (i.e., deciding whether available resources are able to support this fixed structure). Assuming that the structure of a hierarchical scheduler is known *a priori* is quite justified when all applications under that scheduler are part of the same system. It is also justified in small-scale settings in which the number of such applications (and the scale of the infrastructure supporting these applications) is small, and in which the set of applications to be supported is rather static. None of these conditions hold in the emerging practices that fuel the use of virtualization technologies. In such settings, inferring the structure of a feasible, efficient hierarchical scheduler is the challenge addressed in [5] with the use of SLA transformations that were proven to be safe using domain-specific knowledge. In this paper, we showed that domain expertise is not necessary by showing how the use of AARTIFACT could assist the process of deriving SLA transformations.

The idea of transforming task periods for improving schedulability is not new. Related work [26], [27] introduced a period transformation method that involves halving the $C$ and $T$ elements of the periodic task specification. The purpose of this transformation was to increase the priority of a task under RMS. In our work, we consider much more general transformations targeting not only hard, but also soft deadline semantics, and which are possible to derive for overlapping as well as traditional, non-overlapping intervals.

**Machine Verification:** The work presented in this paper illustrates the usefulness of several characteristics of a formal reasoning system that prioritizes usability in its design. The concrete syntax of the system is familiar to the community in question, and it is not necessary to cite explicitly within a formal argument the definitions and propositions that are being employed. These features are recognized as important within several efforts and projects to design systems with similar characteristics [28]–[30]. Furthermore, the system takes advantage of an extensive database of definitions and propositions dealing with common mathematical concepts and provides native support for some of these concepts. This is inspired by work in a subdiscipline of artificial intelligence that deals with the assembly and application of ontologies, such as the Cyc Project [31]. The augmented context integrated into the inference algorithm and its closure operation are similar to structures and algorithms found in work on conjunction closures [15].

There exist few examples of applications of formal reasoning systems within novel research. Some examples include applications within network protocol design [32], cryptography [33], security in computation [34], economic mechanism design [35]. Our work is distinguished from these by our utilization of a system that provides lightweight verification. In this approach, the semantics of the underlying domain is encoded separately by domain experts. The system user is only required to use familiar notation and a supporting interface to encode a correspondence between the application domain and the underlying semantic domain.

VI. CONCLUSION

We have utilized a formal reasoning system to define and reason about a novel framework for rewriting SLAs to enable efficient co-location in a cloud setting. These proofs are accessible to humans and are of a manageable size, demonstrating
the value of employing a user-friendly verification system.

Our on-going research efforts are progressing along two dimensions. Along the first, we are employing AARTIFACT in proving additional conjectured theorems about other models of SLA transformations for workload colocation. We believe additional SLA transformations will enhance the efficiency of our colocation framework (e.g., in [11]), SLA transformations such as those seen in this paper yield a reduction of up to 60% in wasted resources). Along the second, we are interested in using AARTIFACT to reason about other models of SLA “calculus” (e.g., an SLA transformation for cloud storage).

REFERENCES


