

Symbolic Finite State Transducers: Algorithms and Applications

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ABSTRACT

Finite automata and finite transducers are used in a wide range of applications in software engineering, from regular expressions to specification languages. We extend these classic objects with symbolic alphabets represented as parametric theories. Admitting potentially infinite alphabets makes this representation strictly more general and succinct than classical finite transducers and automata over strings. Despite this, the main operations, including composition, checking that a transducer is single-valued, and equivalence checking for single-valued symbolic finite transducers are effective given a decision procedure for the background theory. We provide novel algorithms for these operations and extend composition to symbolic transducers augmented with registers. Our base algorithms are unusual in that they are non-constructive, therefore, we also supply a separate model generation algorithm that can quickly find counterexamples in the case two symbolic finite transducers are not equivalent. The algorithms give rise to a complete decidable algebra of symbolic transducers. Unlike previous work, we do not need any syntactic restriction of the formulas on the transitions, only a decision procedure. In practice we leverage recent advances in satisfiability modulo theory (SMT) solvers. We demonstrate our techniques on four case studies, covering a wide range of applications. Our techniques can synthesize string pre-images in excess of 8,000 bytes in roughly a minute, and we find that our new encodings significantly outperform previous techniques in succinctness and speed of analysis.

1. INTRODUCTION

Finite automata are used in a wide range of applications in software engineering, from regular expressions to specification languages. Nearly every programmer has used a regular expression at one point or another to parse logs or manipulate text. Finite transducers are an extension of finite automata to model functions on lists of elements, which in turn have uses in fields as diverse as computational linguistics and model-based testing. While this formalism is of immense practical use, it suffers from certain drawbacks: in the presence of large alphabets, they can “blow up” in the number of transitions, as each transition can encode only one choice of element from the alphabet. Furthermore, the most common forms cannot handle infinite alphabets.

Symbolic finite transducers are an extension of traditional transducers that attempt to solve these problems by allowing transitions to be labeled with arbitrary formulas in a specified theory. While the concept is straightforward, traditional algorithms for deciding composition, equivalence, and other properties of finite transducers *do not* immediately gener-

alize to the symbolic case. In particular, previous work on symbolic finite transducers have needed to impose syntactic restrictions on formulas to achieve decidable analysis [1, 37]. Our work breaks this barrier and allows for arbitrary formulas from *any decidable background theory*. In practice, we leverage the recent progress in satisfiability modulo theory (SMT) solvers to provide this decision procedure. We find that our algorithms are fast when used with Z3, a state of the art SMT solver.

The restriction we do make is a semantic one: that the symbolic finite transducer is *single-valued*. This restriction is needed because equivalence is undecidable even for standard finite transducers. The single-valuedness property is decidable for symbolic finite transducers. This gives us a way to check transducers arising from practical applications before applying our algorithms.

While it was previously known that equivalence was decidable for single-valued finite transducers, again it does not immediately follow that equivalence should be decidable for single-valued *symbolic* finite transducers because typically even very restricted extensions of finite automata and finite transducers lead to undecidability of the core decision problems. In fact, our proof requires a delicate separation between the “automata theoretic” parts of our algorithms and the use of the decision procedure. Unusually, our algorithm for deciding equivalence is *nonconstructive*: while we can determine that two symbolic finite automata are not equivalent, our proof does not provide a way to find a counterexample. Fortunately, we provide a separate *model generation* semi-decision procedure that can find counterexamples once it is known that two automata are not equivalent.

Figure 1 summarizes known results about finite state transducers (over sequences) and extensions thereof, focusing on the key properties studied in this paper, namely functional *compositionality*, decidability of *equivalence*, and the role of the *alphabet*, in order to place our contributions in a clear context. Section 5 describes previous work, including work on extending automata to trees and streams.

1.1 Applications

Section 4 presents four comprehensive case studies in different areas. Our first case study extends previous work in using symbolic finite transducers to model web “sanitization functions” [4]. Our techniques allow the addition of *register variables*, which enable encoding a sanitizer that could not be handled efficiently by previous symbolic approaches. Our second case study shows an application to analysis of Javascript malware found on the Web. Our third and fourth case studies showcase additional theories beyond strings. Figure 2 summarizes how each case study reflects a novel feature of our work. In addition to these case studies, finite transducers have been employed in other areas such as analysis of web sanitization frameworks, host-based intrusion detection, and natural language processing. Our work immediately applies to these application domains.

1.2 Contributions

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	Effective closure under composition	Equivalence	Alphabet
Finite State Transducers	closed	undecidable in general [22], decidable for single-valued case [42] and finite-valued case [12, 49]	finite set of elements, comparable with equality
Streaming Transducers [1]	closed (for finite alphabets)	decidable	total orders (infinite)
Symbolic Finite Transducers	closed (Proposition 1)	decidable for single-valued case (Theorem 2)	any decidable theory
Symbolic Transducers	closed (extension of Proposition 1)	undecidable (already for the single-valued case, through direct encoding of 2-counter machines)	any decidable theory

Figure 1: Summary of decidability results. The bottom two rows summarize our contributions.

Case study	Section	Feature
HTMLDecode	4.1	ST representation compactness, integer-linear arithmetic
Malware fingerprinting	4.2	SFT composition for program analysis
Image blurring	4.3	non-string module theory
Location privacy	4.4	stream manipulating programs

Figure 2: Summary of case studies.

Our contributions are the following:

- We present novel algorithms for *composition* and *equivalence checking* of symbolic finite transducers. Our algorithms, unlike previous work, make no restrictions on the formulae used in the transducers: we require only a decision procedure for the background theory.
- We show that the single-valuedness property of symbolic transducers is decidable. This gives rise to a decidable complete algebra of symbolic transducers. The impact is that single-valued symbolic transducers can now be “first class” objects for constructing program analyses.
- We present four case studies that demonstrate how our new algorithms enable new applications. We demonstrate experimentally that our algorithms not only terminate, but that they run quickly in practice for problem instances of interest.

1.3 Paper Organization

The rest of this paper is structured as follows. In Section 2 we provide an introduction to symbolic finite state transducers. Section 3 describes the core transducer-based algorithms. Section 4 provides four detailed cases studies of transducer use. Finally, we discuss closely related work in Section 5 and conclude in Section 6.

2. SYMBOLIC FINITE TRANSUCERS

We now formally define symbolic finite transducers, we give examples of how these objects model program behavior, and we define analyses that may be conducted on such transducers. We assume a *background theory* that is typed. We write σ and γ for given types, and we write \mathcal{U}^σ or Σ (resp. \mathcal{U}^γ or Γ) for the corresponding domain of elements of type σ (resp. γ). The intuition behind the notation throughout the paper is that Σ is a domain of individual input elements and Γ is a domain of individual output elements. Terms and formulas are defined as usual over the language of the background theory and are assumed to be well-typed.

A σ -predicate φ is a formula representing a subset $[\varphi]$ of Σ . $\mathcal{P}(\sigma)$ denotes a given set of σ -predicates such that, for

each element $a \in \Sigma$ there is a predicate representing $\{a\}$, **t** (true), and **f** (false) are in $\mathcal{P}(\sigma)$, and $\mathcal{P}(\sigma)$ is effectively closed under Boolean operations.

A σ/γ -term f is a term representing a function $\llbracket f \rrbracket$ from Σ to Γ . We write $\mathcal{F}(\sigma \rightarrow \gamma)$ for a given set of σ/γ -terms and assume that all $b \in \Gamma$ are also constants in $\mathcal{F}(\sigma \rightarrow \gamma)$, i.e., $\llbracket b \rrbracket(a) = b$ for all $a \in \Sigma$.

A *label theory*, denoted $\mathcal{P}(\sigma)/\mathcal{F}(\sigma \rightarrow \gamma)$, contains all formulas φ and $\varphi(x) \wedge f(x) \neq g(x)$ for $\varphi \in \mathcal{P}(\sigma)$ and $f, g \in \mathcal{F}(\sigma \rightarrow \gamma)$. A theory Ψ is *decidable* when satisfiability for $\varphi \in \Psi$, denoted $IsSat(\varphi)$, is decidable. We assume an effective *witness* function for a σ -predicate φ such that, if $IsSat(\varphi)$ then $witness(\varphi) \in \llbracket \varphi \rrbracket$.

Two σ/γ -terms f and g are *equivalent relative* to a σ -predicate φ , denoted $f \equiv_\varphi g$, when $\varphi(x) \wedge f(x) \neq g(x)$ is unsatisfiable.

Example 1: An example of a decidable label theory is quantifier-free linear integer arithmetic with at most one fixed free variable x for the input element. \square

Throughout the paper we assume that σ -predicates are formulas with (at most) one fixed free variable x of type σ , and σ/γ -terms are terms of type γ that have (at most) one fixed free variable x of type σ for the input. Given a set X , we write X^* for the *Kleene closure* of X .

Next, we describe an extension of finite state transducers through a symbolic representation of labels. The advantage of the extension is succinctness and modularity with respect to any given label theory. It naturally separates the finite state transition graph from the label theory.

Definition 1: A Symbolic Finite Transducer (SFT) over $\mathcal{P}(\sigma)/\mathcal{F}(\sigma \rightarrow \gamma)$ is a tuple (Q, q^0, F, R) , where Q is a finite set of states, $q^0 \in Q$ is the initial state, $F \subseteq Q$ is the set of final states, and $R \subseteq Q \times \mathcal{P}(\sigma) \times (\mathcal{F}(\sigma \rightarrow \gamma))^* \times Q$ is a finite set of rules. \square

We use the more intuitive notation $p \xrightarrow{\varphi/v} q$ for a rule $(p, \varphi, v, q) \in R_A$ and call φ its *guard*. We omit the index A when A is clear from the context. Although $v \in (\mathcal{F}(\sigma \rightarrow \gamma))^*$ is a finite sequence of σ/γ -terms, $v = [f_0, \dots, f_{|v|-1}]$ denotes a function $\llbracket v \rrbracket$ from Σ to Γ^* such that, for $a \in \Sigma$,

$$\llbracket v \rrbracket(a) = [\llbracket f_0 \rrbracket(a), \dots, \llbracket f_{|v|-1} \rrbracket(a)].$$

We lift the definition of relative equivalence of σ/γ -terms to sequences v and w of σ/γ -terms: $v \equiv_\varphi w$ iff $|v| = |w|$ and for all i , $0 \leq i < |v|$, $v[i] \equiv_\varphi w[i]$. We use the notation of rules to also denote concrete transitions when the intension is clear from the context. For $p, q \in Q_A$, $a \in \Sigma$ and $\mathbf{v} \in \Gamma^*$:

$$p \xrightarrow{a/\mathbf{v}} q \stackrel{\text{def}}{=} \exists \varphi v (p \xrightarrow{\varphi/v} q \wedge a \in \llbracket \varphi \rrbracket \wedge \mathbf{v} = \llbracket v \rrbracket(a)),$$

i.e., the transition $p \xrightarrow{a/\mathbf{v}} q$ is an *instance* of a rule. Con-

catenation of two sequences u and v is denoted $u \cdot v$.

Definition 2: For $\mathbf{u} \in \Sigma^*$ and $\mathbf{v} \in \Gamma^*$, $p \xrightarrow{\mathbf{u}/\mathbf{v}}_A q$ denotes the reachability relation: there exists a path of transitions from p to q in A with input \mathbf{u} and output \mathbf{v} : let $n = |\mathbf{u}| - 1$,

$$p = p_0 \xrightarrow{\mathbf{u}[0]/\mathbf{v}_0} p_1 \xrightarrow{\mathbf{u}[1]/\mathbf{v}_1} p_2 \cdots p_n \xrightarrow{\mathbf{u}[n]/\mathbf{v}_n} p_{n+1} = q$$

such that $\mathbf{v} = \mathbf{v}_0 \cdot \mathbf{v}_1 \cdots \mathbf{v}_{n-1}$. Let $p \xrightarrow{\epsilon/\epsilon}_A p$ for $p \in Q_A$. \boxtimes

Definition 3: The transduction of A , denoted \mathcal{T}_A , is the following function from Σ^* to 2^{Γ^*} :

$$\mathcal{T}_A(\mathbf{u}) \stackrel{\text{def}}{=} \{\mathbf{v} \in \Gamma^* \mid \exists q \in F_A (q_A \xrightarrow{\mathbf{u}/\mathbf{v}} q)\}.$$

Equivalently, \mathcal{T}_A is viewed as the binary relation, or subset of $\Sigma^* \times \Gamma^*$, such that $\mathcal{T}_A(\mathbf{u}, \mathbf{v})$ iff $\mathbf{v} \in \mathcal{T}_A(\mathbf{u})$. The domain of A , $\mathcal{D}(A) \stackrel{\text{def}}{=} \{\mathbf{u} \in \Sigma^* \mid \mathcal{T}_A(\mathbf{u}) \neq \emptyset\}$. \boxtimes

The following subclass of SFTs captures transductions that behave as partial functions from Σ^* to Γ^* .

Definition 4: A is single-valued when $|\mathcal{T}_A(\mathbf{u})| \leq 1$ for all $\mathbf{u} \in \Sigma^*$. \boxtimes

A sufficient condition for single-valuedness is determinism.

Definition 5: A is deterministic when, for all $p \xrightarrow{\varphi/\psi}_A q$ and $p \xrightarrow{\varphi'/\psi'}_A r$, if $\text{IsSat}(\varphi \wedge \psi)$ then $q = r$ and $v \equiv_{\varphi \wedge \psi} w$. \boxtimes

In terms of concrete transitions, determinism of A means that if $p \xrightarrow{\mathbf{a}/\mathbf{v}}_A q$ and $p \xrightarrow{\mathbf{a}'/\mathbf{w}}_A r$ then $(\mathbf{v}, q) = (\mathbf{w}, r)$. It follows by induction over $|\mathbf{u}|$ for $\mathbf{u} \in \Sigma^*$ that, if $p \xrightarrow{\mathbf{u}/\mathbf{v}}_A q$ then (\mathbf{v}, q) is unique for the given p and \mathbf{u} , in particular when $p = q_A^0$ and $q \in F_A$, and thus A is single-valued. Determinism is, however, not a necessary condition for single-valuedness, as is illustrated below. In the following examples, all SFTs are single-valued. The first example illustrates a few simple functional list transformations, expressed as deterministic SFTs that illustrate how global properties of SFTs depend on the theory of labels.

Example 2: Let the input type and the output type be \mathbb{Z} . All SFTs have a single state here. Predicates and terms are terms in integer linear arithmetic. *Negate* multiplies all elements by -1. *Increment* adds 1 to each element. *DeleteZeros* deletes all zeros from the input.

$$\begin{aligned} R_{\text{Negate}} &= \{p \xrightarrow{t/[-x]} p\} \\ R_{\text{Increment}} &= \{q \xrightarrow{t/[1+x]} q\} \\ R_{\text{DeleteZeros}} &= \{r \xrightarrow{x=0/[]} r, \quad r \xrightarrow{x \neq 0/[x]} r\} \end{aligned}$$

Properties such as commutativity and idempotence of SFTs depend on the theory of labels. For example, whether *Negate* and *DeleteZeros* commute or whether *DeleteZeros* is idempotent depend on properties of integer addition and multiplication. None of the examples can be expressed as traditional finite state transducers over a finite alphabet. Our results about composition and equivalence checking, that are discussed below, allow us to effectively establish such properties modulo decidability of a given label theory. \boxtimes

The following example illustrates a common string transformation where the use of nondeterministic SFTs is essential.

Example 3: Suppose that the following C# code is intended to implement a function `GetTags` that extracts from a given input stream of characters all substrings of the form

$[\langle', x, \rangle']$, where $x \neq \langle'$. For example

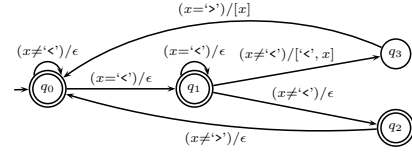
$$\text{GetTags}("\langle\langle s \rangle\rangle \langle f \rangle \langle t \rangle") = "\langle s \rangle \langle f \rangle"$$

```

1: int q = 0; char c = (char)0;
2: foreach (char x in input) {
3:   if (q == 0) { if (x == '<') q = 1; else q = 0;}
4:   else if (q == 1) { if (x == '<') q = 1; else q = 2;}
5:   else if (q == 2) { if (x == '>') { yield return '<';
7:                                     yield return c;
8:                                     yield return '>';}
9:   c = x; }

```

Note that the variable q keeps track of the relative position in the pattern $[\langle', x, \rangle']$ and c records the previous character. The corresponding single-valued SFT *GetTags* is:



GetTags is nondeterministic because there are two rules from q_1 to q_2 and q_3 , respectively, yielding different outputs for the same input. If the characters are represented as integers, then a deterministic version of *GetTags* does not exist, and if the characters are represented as 16-bit bitvectors (that corresponds precisely to the standard UTF-16 encoding of characters in C#) then the size of the equivalent deterministic SFT is 2^{16} times larger. (Example 7 below explains how *GetTags* is constructed from the C# code.) \boxtimes

3. SFT ALGORITHMS

In this section we study algorithms for *composition* and *equivalence* of SFTs. First, we show that SFTs are effectively closed under composition. Next, we provide an efficient algorithm for *single-valued equality* of SFTs modulo a decidable theory of labels. Finally we introduce an algebra of SFTs that enables a variety of practically useful decision problems, such as deciding single-valuedness, and deciding commutativity and idempotence of single-valued SFTs.

3.1 Composition of SFTs

Given two transductions \mathcal{T}_1 and \mathcal{T}_2 , $\mathcal{T}_1 \circ \mathcal{T}_2$ denotes the following function:

$$\mathcal{T}_1 \circ \mathcal{T}_2 \stackrel{\text{def}}{=} \lambda \mathbf{u}. \bigcup_{\mathbf{v} \in \mathcal{T}_1(\mathbf{u})} \mathcal{T}_2(\mathbf{v}).$$

This definition follows the convention in [20]. Notice that \circ applies first \mathcal{T}_1 , then \mathcal{T}_2 , contrary to how \circ is used for standard function composition. Note also that single-valuedness is trivially preserved by composition.

We say that $\mathcal{P}(\sigma)/\mathcal{F}(\sigma \rightarrow \delta)$ and $\mathcal{P}(\delta)/\mathcal{F}(\delta \rightarrow \gamma)$ are *composable* if $\mathcal{P}(\sigma)/\mathcal{F}(\sigma \rightarrow \gamma)$ is a label theory such that:

- if $f \in \mathcal{F}(\sigma \rightarrow \delta)$, $g \in \mathcal{F}(\delta \rightarrow \gamma)$ then $g(f) \in \mathcal{F}(\sigma \rightarrow \gamma)$ such that $\llbracket g(f) \rrbracket(a) = \llbracket g \rrbracket(\llbracket f \rrbracket(a))$.
- if $\varphi \in \mathcal{P}(\delta)$ and $f \in \mathcal{F}(\sigma \rightarrow \delta)$ then $\varphi(f) \in \mathcal{P}(\sigma)$ such that $a \in \llbracket \varphi(f) \rrbracket \Leftrightarrow \llbracket f \rrbracket(a) \in \llbracket \varphi \rrbracket$.

Proposition 1: Let A and B be SFTs over composable label theories. Then there exists an SFT $A \circ B$ that is obtained effectively from A and B such that $\mathcal{T}_{A \circ B} = \mathcal{T}_A \circ \mathcal{T}_B$.

The algorithm for $A \circ B$ can be efficiently implemented with a DFS procedure that, by assuming decidability of $\mathcal{P}(\sigma)$, eliminates incrementally all composed rules that have unsatisfiable guards and finally eliminates all *deadends* (*deadlock states*: states from which no final state is reachable).

3.2 Equivalence of SFTs

We introduce an algorithm for deciding equivalence of *single-valued* SFTs. While general equivalence of finite state transducers is undecidable [22] the undecidability is caused by allowing unboundedly many different outputs for a given input. Single-valued transducers, furthermore, correspond closely to functional transformations over lists computed by concrete programs. As illustrated above, this does not (in general) rule out nondeterministic SFTs.

SFTs A and B are *equivalent*, $A \equiv B$, when $\mathcal{T}_A = \mathcal{T}_B$. Deciding $A \equiv B$ reduces to two independent tasks:

Domain equivalence : $\mathcal{D}(A) = \mathcal{D}(B)$.

Partial equivalence : $(\forall \mathbf{v} \in \mathcal{D}(A) \cap \mathcal{D}(B)) \mathcal{T}_A(\mathbf{v}) = \mathcal{T}_B(\mathbf{v})$

A *symbolic finite automaton* or *SFA* is an SFT all of whose outputs are ϵ . Let $\mathbf{d}(A)$ denote the SFA obtained from the SFT A by replacing all outputs by ϵ . Then $\mathcal{D}(A) = \mathcal{D}(B)$ iff $\mathbf{d}(A) \equiv \mathbf{d}(B)$. Equivalence of SFAs is decidable provided that satisfiability for $\mathcal{P}(\sigma)$ is decidable [47] (the decidability of SFA equivalence depends on the assumption that $\mathcal{P}(\sigma)$ is closed under complementation).

For developing a decision procedure for partial equivalence of single-valued SFTs we use the following weak form of partial equivalence.

Single-valued equality (or 1-equality $A \stackrel{\perp}{=} B$) :

$$\forall \mathbf{u} \mathbf{v} \mathbf{w} ((\mathbf{v} \in \mathcal{T}_A(\mathbf{u}) \wedge \mathbf{w} \in \mathcal{T}_B(\mathbf{u})) \Rightarrow \mathbf{v} = \mathbf{w})$$

Proposition 2: A is single-valued iff $A \stackrel{\perp}{=} A$. If A and B are single-valued then $A \stackrel{\perp}{=} B$ iff A and B are partially equivalent.

Single-valued equality of two SFTs A and B may fail for two reasons. There is an input $\mathbf{u} \in \mathcal{D}(A) \cap \mathcal{D}(B)$ and outputs $\mathbf{v} \in \mathcal{T}_A(\mathbf{u})$ and $\mathbf{w} \in \mathcal{T}_B(\mathbf{u})$ such that:

1. A has a *length-conflict* with B : $|\mathbf{v}| \neq |\mathbf{w}|$.
2. A has a *position-conflict* with B : $|\mathbf{v}| = |\mathbf{w}|$ and, for some position i , $0 \leq i < |\mathbf{v}|$, $\mathbf{v}[i] \neq \mathbf{w}[i]$.

We introduce the following basic product construction of SFTs as a generalization of the product of SFAs [47]. The product construction is most effectively realized by using a DFS procedure. Note that the product of SFTs is a “2-output-SFT” (SFTs are not closed under product).

Definition 6: The product of SFTs A and B , denoted $A \times B$, is defined as the least fixpoint of pair states $Q \subseteq Q_A \times Q_B$ and rules under the following conditions:

- $(q_A^0, q_B^0) \in Q$,
- if $(p_1, p_2) \in Q$, $p_1 \xrightarrow{\varphi/u} q_1$, and $p_2 \xrightarrow{\psi/v} q_2$, then
 - $(q_1, q_2) \in Q$ and
 - $(p_1, p_2) \xrightarrow{\varphi \wedge \psi / (u, v)}_{A \times B} (q_1, q_2)$,

provided that $\text{IsSat}(\varphi \wedge \psi)$.

All deadends, noninitial states from which $F_A \times F_B$ is not reachable, are eliminated from $A \times B$. \square

Example 4: Consider the SFT *GetTags* in Example 3. Then $\text{GetTags} \times \text{GetTags}$ has rules $(p, p) \xrightarrow{\varphi/(u, u)} (q, q)$ for all $p \xrightarrow{\varphi/u}_{\text{GetTags}} q$ and no other rules due to elimination of deadends, e.g., (q_3, q_2) is a deadend because the guard of the only possible rule $(q_3, q_2) \xrightarrow{x='>\wedge x \neq '>' / ([x], \epsilon)} (q_0, q_0)$ from (q_3, q_2) is unsatisfiable. \square

Let $\mathcal{D}(A \times B)$ denote the set of inputs that are accepted by the product. It follows from the product construction that:

$$\mathcal{D}(A \times B) = \mathcal{D}(A) \cap \mathcal{D}(B). \quad (1)$$

The reachability relation is lifted to $A \times B$ and the following holds for $p = (p_1, p_2), q = (q_1, q_2) \in Q_{A \times B}$:

$$p \xrightarrow{\mathbf{u}/(\mathbf{v}, \mathbf{w})}_{A \times B} q \Leftrightarrow p_1 \xrightarrow{\mathbf{u}/\mathbf{v}}_{A} q_1 \wedge p_2 \xrightarrow{\mathbf{u}/\mathbf{w}}_{B} q_2. \quad (2)$$

We omit the index $A \times B$ when it is clear from the context. For $q \in Q_{A \times B}$, and $k \geq 0$, define the *offset relation*,

$$q \Delta k \stackrel{\text{def}}{=} \exists \mathbf{u} \mathbf{v} \mathbf{w} (q_{A \times B}^0 \xrightarrow{\mathbf{u}/(\mathbf{v}, \mathbf{w})} q \wedge k = |\mathbf{v}| - |\mathbf{w}|) \quad (3)$$

The intuition behind $q \Delta k$ is that, for some common input \mathbf{u} , there exists an output \mathbf{v} from A that is either ahead of an output \mathbf{w} from B with k -positions at product state q when $k > 0$, or behind when $k < 0$. The following lemma is used to detect length-conflicts.

Lemma 1: If there exists $q \in Q_{A \times B}$ and $m \neq n$ such that $q \Delta m$ and $q \Delta n$ then A has a length-conflict with B . Moreover, A has a length-conflict with B iff there exists $q \in F_{A \times B}$ and $m \neq 0$ such that $q \Delta m$.

Lemma 1 suggests an efficient DFS algorithm to detect if a length-conflict exists and otherwise computes the fixed offset $\text{offs}(p)$ such that $p \Delta \text{offs}(p)$. In order to decide if a position-conflict exists between A and B , we first assume that A and B have no length-conflicts and assume that $\text{offs}(p)$ is defined and $\text{offs}(p) = 0$ for $p \in F_{A \times B}$. We say that $(\alpha, \beta) \in \Gamma^* \times \Gamma^*$ is a *promise* of a product state $p \in Q_{A \times B}$ when the following holds:

$$(\alpha = \epsilon \vee \beta = \epsilon) \wedge \exists \mathbf{u} \mathbf{v} \mathbf{w} (|\mathbf{v}| = |\mathbf{w}| \wedge q_{A \times B}^0 \xrightarrow{\mathbf{u}/(\mathbf{v} \cdot \alpha, \mathbf{w} \cdot \beta)} p)$$

It follows that

$$(|\alpha|, |\beta|) = \begin{cases} (\text{offs}(p), 0), & \text{if } \text{offs}(p) \geq 0; \\ (0, -\text{offs}(p)), & \text{otherwise.} \end{cases}$$

Lemma 2: If $A \stackrel{\perp}{=} B$ then each product state in $Q_{A \times B}$ has a fixed promise.

Proof. Assume $A \stackrel{\perp}{=} B$. Suppose, by way of contradiction, that there exists $p \in Q_{A \times B}$ with two distinct promises (α, β) and (α', β') . By Lemma 1 we know that $|\alpha| = |\alpha'|$ and $|\beta| = |\beta'|$ and either $\alpha = \epsilon$ or $\beta = \epsilon$. Suppose $\beta = \epsilon$ (the case $\alpha = \epsilon$ is symmetrical). Thus $\alpha \neq \alpha'$ (or else the promises are identical). By definition of promises there exist $\mathbf{u}, \mathbf{v}, \mathbf{w}, \mathbf{u}', \mathbf{v}', \mathbf{w}'$ such that, $|\mathbf{v}| = |\mathbf{w}|$, $|\mathbf{v}'| = |\mathbf{w}'|$,

$$q_{A \times B}^0 \xrightarrow{\mathbf{u}/(\mathbf{v} \cdot \alpha, \mathbf{w})} p, \quad q_{A \times B}^0 \xrightarrow{\mathbf{u}'/(\mathbf{v}' \cdot \alpha', \mathbf{w}')} p.$$

Since p is not a deadend, there exist $\mathbf{u}'', \mathbf{v}'', \mathbf{w}''$ and $q^f \in F_{A \times B}$ such that

$$p \xrightarrow{\mathbf{u}''/(\mathbf{v}'', \mathbf{w}'')} q^f.$$

It follows from $A \stackrel{\perp}{=} B$ that

$$\left. \begin{array}{l} \mathbf{v} \cdot \alpha \cdot \mathbf{v}'' \in \mathcal{T}_A(\mathbf{u} \cdot \mathbf{u}'') \\ \mathbf{w} \cdot \mathbf{w}'' \in \mathcal{T}_B(\mathbf{u} \cdot \mathbf{u}'') \end{array} \right\} \Rightarrow \mathbf{v} \cdot \alpha \cdot \mathbf{v}'' = \mathbf{w} \cdot \mathbf{w}''$$

$$\left. \begin{array}{l} \mathbf{v}' \cdot \alpha' \cdot \mathbf{v}'' \in \mathcal{T}_A(\mathbf{u}' \cdot \mathbf{u}'') \\ \mathbf{w}' \cdot \mathbf{w}'' \in \mathcal{T}_B(\mathbf{u}' \cdot \mathbf{u}'') \end{array} \right\} \Rightarrow \mathbf{v}' \cdot \alpha' \cdot \mathbf{v}'' = \mathbf{w}' \cdot \mathbf{w}''$$

and, since $|\mathbf{v}| = |\mathbf{w}|$ and $|\mathbf{v}'| = |\mathbf{w}'|$, it follows that $\alpha \cdot \mathbf{v}'' = \mathbf{w}''$ and $\alpha' \cdot \mathbf{v}'' = \mathbf{w}''$, contradicting that $\alpha \neq \alpha'$. \square

Similar to Lemma 1, Lemma 2 suggests an efficient DFS algorithm to detect if distinct promises exists for some product state and otherwise computes the fixed promise $\text{prom}(p)$ for all $p \in Q_{A \times B}$.

Example 5: The product $\text{GetTags} \times \text{GetTags}$ in Example 4 has trivially no length-conflicts because offsets of all product states are 0 and thus promises of all product states are (ϵ, ϵ) . \boxtimes

Finally, assuming $A \times B$ has no length-conflicts and each product state $p \in Q_{A \times B}$ has a fixed promise $\text{prom}(p) = (\alpha, \beta)$, then p is *conflict-free*, when, for all rules $p \xrightarrow{\varphi/(v,w)} q$, the maximal prefixes of $\alpha \cdot v$ and $\beta \cdot w$ are equivalent relative to φ : let $k = \min(|\alpha \cdot v|, |\beta \cdot w|) - 1$,

$$\bigwedge_{j=0}^k (\forall a(a \in \llbracket \varphi \rrbracket \Rightarrow \llbracket \alpha \cdot v[j] \rrbracket(a) = \llbracket \beta \cdot w[j] \rrbracket(a))). \quad (4)$$

We say that p is a *conflict-state* if p is not conflict-free. Verifying absense of conflict-states is a linear search over rules in $A \times B$ that verifies the condition (4) for each rule.

Lemma 3: If $A \times B$ has fixed promises and no length-conflicts then $A \stackrel{\perp}{\neq} B \Leftrightarrow Q_{A \times B}$ contains a conflict-state.

Proof. Assume that $A \times B$ is as stated.

(\Leftarrow): existence of a conflict-state p implies, by definition, that there exists a position-conflict, since p is both reachable and not a deadend.

(\Rightarrow): Assume $A \stackrel{\perp}{\neq} B$. We show that there exists a conflict-state. Since A and B have no length-conflicts there exist $\mathbf{u}, \mathbf{v}, \mathbf{w}$ such that $\mathbf{v} \in \mathcal{T}_A(\mathbf{u})$, $\mathbf{w} \in \mathcal{T}_B(\mathbf{u})$, $|\mathbf{v}| = |\mathbf{w}|$, and there exists a position i , $0 \leq i < |\mathbf{v}|$, such that $\mathbf{v}[i] \neq \mathbf{w}[i]$. Fix i to be the smallest such position. So there exists $\mathbf{u}_1, a, \mathbf{v}_1, \mathbf{v}_2, \mathbf{w}_2, \alpha, \beta, p, q$ such that $\mathbf{u}_1 \cdot a$ is a prefix of \mathbf{u} , $\mathbf{v}_1 \cdot \alpha \cdot \mathbf{v}_2$ is a prefix of \mathbf{v} , $\mathbf{v}_1 \cdot \beta \cdot \mathbf{w}_2$ is a prefix of \mathbf{w} , and

$$q_{A \times B}^0 \xrightarrow{\mathbf{u}_1 / (\mathbf{v}_1 \cdot \alpha \cdot \mathbf{v}_1 \cdot \beta)} p \xrightarrow{\alpha / (\mathbf{v}_2 \cdot \mathbf{w}_2)} q$$

where $\text{prom}(p) = (\alpha, \beta)$ and $(\alpha \cdot \mathbf{v}_2)[j] \neq (\beta \cdot \mathbf{w}_2)[j]$ with $j = i - |\mathbf{v}_1|$. So there exists a rule $p \xrightarrow{\varphi / (\mathbf{v}_2 \cdot \mathbf{w}_2)} q$ such that $a \in \llbracket \varphi \rrbracket$ but $\llbracket (\alpha \cdot \mathbf{v}_2)[j] \rrbracket(a) \neq \llbracket (\beta \cdot \mathbf{w}_2)[j] \rrbracket(a)$. Thus, p is a conflict-state because (4) does not hold. \square

The following theorem describes precisely the assumptions under which 1-equality of SFTs is decidable.

Theorem 1 (SFT-1-equality): If A and B are SFTs over a decidable label theory then $A \stackrel{\perp}{=} B$ is decidable. Moreover, if the complexity of the label theory for instances of size m is $f(m)$ then the complexity of $A \stackrel{\perp}{=} B$ is $O(n^2 \cdot f(m))$ where n is the number of rules and m the size of the rules.

Proof. By using Lemmas 1, 2 and 3. Deciding satisfiability of φ is needed in the construction of $A \times B$. Deciding $f \equiv_{\varphi} g$ is needed for deciding validity of the formula (4). In Lemma 2, we need to decide if f is constant relative to a satisfiable formula φ : decide if $f \equiv_{\varphi} \llbracket f \rrbracket(\text{witness}(\varphi))$.

Lemmas 1, 2 and 3 are combined into a single DFS algorithm, shown in Figure 3, that decides 1-equality of SFTs over a decidable label theory. Line 6 corresponds to detec-

```

Decide1equality( $A, B$ )  $\stackrel{\text{def}}{=}
1 \ C := A \times B; Q := \{q_C^0 \mapsto (\epsilon, \epsilon)\}; S := \text{stack}(q_C^0);
2 \ \text{while } S \neq \emptyset
3 \ \ \ p := \text{pop}(S); (\alpha, \beta) := Q(p);
4 \ \ \ \text{foreach } (p, \varphi, (u_1, v_1), q) \in R_C(p)
5 \ \ \ \ \ (u, v) := (\alpha \cdot u_1, \beta \cdot v_1);
6 \ \ \ \ \ \text{if } q \in F_C \wedge |u| \neq |v| \ \text{return } \mathbf{f};
7 \ \ \ \ \ \text{if } |u| \geq |v|
8 \ \ \ \ \ \ \ \ \ \text{if } \bigvee_{i=0}^{|v|-1} u[i] \neq_{\varphi} v[i] \ \text{return } \mathbf{f};
9 \ \ \ \ \ \ \ \ \ \ w := [u[|v|], \dots, u[|u| - 1]]; \mathbf{w} := \llbracket w \rrbracket(\text{witness}(\varphi));
10 \ \ \ \ \ \ \ \ \ \ \text{if } w \neq_{\varphi} \mathbf{w} \vee (q \in \text{Dom}(Q) \wedge Q(q) \neq (\mathbf{w}, \epsilon)) \ \text{return } \mathbf{f};
11 \ \ \ \ \ \ \ \ \ \ \text{if } q \notin \text{Dom}(Q) \ \text{push}(q, S); Q(q) := (\mathbf{w}, \epsilon);
12 \ \ \ \ \ \ \ \ \ \ \text{if } |u| < |v| \dots \text{(symmetrical to the case } |u| \geq |v|)
13 \ \text{return } \mathbf{t};$ 
```

Figure 3: 1-equality algorithm for SFTs.

tion of a final state with non-zero offset by using Lemma 1. Line 8 corresponds to use of Lemma 3(\Leftarrow). Line 10 corresponds to use of Lemma 2. Line 13 corresponds to use of contraposition of Lemma 3(\Rightarrow).

The number of iterations of the loop as well as in the product construction is bounded by $|R_A| \cdot |R_B|$. The algorithm uses satisfiability checks during product construction in line 1, and in the loop in lines 8 and 10. In line 8 the number of checks is linear in the length of the output sequence v : decide if there exists i , $0 \leq i < |v|$, and $\varphi(x) \wedge u[i](x) \neq v[i](x)$ is satisfiable. Similarly for line 10. The complexity follows. \square

Theorem 1 shows that complexity of 1-equality of SFTs depends on the complexity of the label theory. For example, if we use linear arithmetic with one free variable as the label theory, and guards are represented in normalized form as conjunctions of linear inequalities, then the Fourier-Motzkin elimination procedure [14] implies a polynomial worst-case complexity of 1-equality.

3.3 Algebra of SFTs

We introduce an algebra of SFTs, in Figure 4, that allows us to express several useful decision problems involving SFTs and SFAs. Note that $B \circ A$ of an SFT B with an SFA A is again an SFA because all the outputs of $B \circ A$ are empty. We call $B \circ A$ the *inverse image of B under A* . The definition of $B \upharpoonright A$ in our algebra is as follows.

Definition 7: Let B be an SFT and A an SFA. The domain restriction of B for A , denoted $B \upharpoonright A$, is the SFT obtained

$$\begin{array}{ll}
\sigma, \delta, \gamma & ::= \text{types} \\
\text{sfa}^{\sigma} & ::= \text{explicit dfn of an SFA over } \mathcal{P}(\sigma) \\
\text{sft}^{\sigma/\gamma} & ::= \text{explicit dfn of an SFT over } \mathcal{P}(\sigma)/\mathcal{F}(\sigma \rightarrow \gamma) \\
A^{\sigma} & ::= \text{sfa}^{\sigma} \mid A^{\sigma} - A^{\sigma} \mid A^{\sigma} \times A^{\sigma} \mid B^{\sigma/\gamma} \circ A^{\gamma} \\
B^{\sigma/\gamma} & ::= \text{sft}^{\sigma/\gamma} \mid B^{\sigma/\delta} \circ B^{\delta/\gamma} \mid B^{\sigma/\gamma} \upharpoonright A^{\sigma} \\
F & ::= A^{\sigma} \subseteq A^{\sigma} \mid B^{\sigma/\gamma} \stackrel{\perp}{=} B^{\sigma/\gamma} \mid F \wedge F \mid \neg F
\end{array}$$

Figure 4: Algebra of SFTs; A is a valid SFA expression; B is a valid SFT expression; F is a valid formula; $\mathcal{P}(\sigma)/\mathcal{F}(\sigma \rightarrow \delta)$ and $\mathcal{P}(\delta)/\mathcal{F}(\delta \rightarrow \gamma)$ are assumed composable, the composition is $\mathcal{P}(\sigma)/\mathcal{F}(\sigma \rightarrow \gamma)$.

from $B \times A$ by eliminating the second output component ϵ from all the rules. \square

The following property follows from (1) and (2).

$$\mathcal{T}_{B \upharpoonright A}(\mathbf{u}) = \begin{cases} \mathcal{T}_B(\mathbf{u}), & \text{if } \mathbf{u} \in \mathcal{D}(A); \\ \emptyset, & \text{otherwise.} \end{cases} \quad (5)$$

We say that the SFT algebra in Figure 4 is *decidable* if validity of all the formulas F in the algebra is decidable.

Theorem 2 (SFT-algebra): The algebra of SFTs is decidable if the label theories are decidable.

Proof. The SFA operations are effectively closed under intersection an complement and equivalence is decidable if satisfiability of the guards is decidable [47]. Decidability of 1-equality of SFTs is Theorem 1. Closure under composition is Proposition 1. Domain restriction is given in (5). \square

The following corollary identifies a collection of practically relevant decision problems that follow from Theorem 2. *Subsumption* of SFTs, $A \sqsubseteq B$, is the problem of deciding if $\mathcal{T}_A(\mathbf{u}) \subseteq \mathcal{T}_B(\mathbf{u})$ for all \mathbf{u} . *Reachability* is the problem of existence of an input that is transformed to an output accepted by an SFA.

Corollary 1: The following decision problems over single-valued SFTs over a decidable label theory are decidable: Subsumption; Equivalence; Idempotence; Commutativity; Reachability.

Proof. Assume A and B are single-valued SFTs and recall Proposition 2. Subsumption, $A \sqsubseteq B$, is $\mathbf{d}(A) \subseteq \mathbf{d}(B) \wedge A \perp B$. Equivalence, $A \equiv B$, is $A \sqsubseteq B \wedge B \sqsubseteq A$. Idempotence is $A \equiv A \circ A$. Commutativity is $A \circ B \equiv B \circ A$. Reachability of a given output SFA D is $A \circ D \neq \emptyset$. \square

The following example illustrates a use of the SFT algebra for reachability analysis of SFTs. The example is a digest behind security analysis of string sanitizers with respect to known XSS attack vectors.

Example 6: Consider the SFT $B = \text{GetTags}$ from Example 3. Is it possible that B does not detect all tags? In other words, does there exist an input \mathbf{u} that matches the regex $P = ". * \langle [^<] \rangle . *"$ but $\mathcal{T}_B(\mathbf{u}) = \{\epsilon\}$?

Let A_ϵ and A_\emptyset be the SFAs such that $\mathcal{D}(A_\epsilon) = \{\epsilon\}$ and $\mathcal{D}(A_\emptyset) = \emptyset$. Let A_P be the SFA that accepts all strings that match P .

The question is equivalent to deciding if (6) fails,

$$\underbrace{(B \upharpoonright A_P) \circ A_\epsilon}_{D} \equiv A_\emptyset \quad (6)$$

because, for $\mathbf{u} \in \Sigma^*$,

$$\begin{aligned} \mathbf{u} \in \mathcal{D}(D) &\Leftrightarrow \exists \mathbf{v} (\mathcal{T}_{B \upharpoonright A_P}(\mathbf{u}) = \{\mathbf{v}\} \wedge \mathbf{v} \in \mathcal{D}(A_\epsilon)) \\ &\Leftrightarrow \mathcal{T}_{B \upharpoonright A_P}(\mathbf{u}) = \{\epsilon\} \\ &\Leftrightarrow \mathcal{T}_B(\mathbf{u}) = \{\epsilon\} \wedge \mathbf{u} \in \mathcal{D}(A_P) \end{aligned}$$

It turns out that D (when minimized) is an SFA with 8 states, e.g., " $\langle \mathbf{a} \rangle$ " $\in \mathcal{D}(D)$. (What is remarkable is that D can be effectively converted back to a regex that describes all inputs where tags are not detected.) By considering the witness, it can easily be traced back to the missing case in the `GetTags` program: line 8 of the C# code should be `q = (x == '<' ? 1 : 0);`. By verifying (6) for the SFT corresponding to the fixed code, we can verify the new code indeed detects all tags. \square

It also follows from Theorem 1 and Proposition 2, that we can decide single-valuedness of SFTs. This is a practically valuable result that lifts the burden of the semantic assumption of single-valuedness in decision problems that assume single-valuedness.

Corollary 2: Single-valuedness of SFTs over a decidable label theory is decidable.

3.4 Extension with Registers

We extend SFTs to symbolic transducers or STs by allowing the use of registers. This will provide a more succinct representation, and enable more efficient symbolic analysis methods to be used, by taking advantage of recent advances in SMT technology [15]. An ST uses a set of variables called *registers* as a symbolic representation of states. The rules of an ST are guarded commands with a symbolic input and output component. Since the finite state component of an SFT can be represented with a particular register of finite type, the explicit state component is omitted from STs. Moreover, by using Cartesian product types, we represent multiple registers with a single (compound) register.

Definition 8: A Symbolic Transducer or ST with input type σ , output type γ and register type ρ is a tuple (q^0, ϑ, R) , where $q^0 \in \mathcal{U}^\rho$ is the initial state, $\vartheta \in \mathcal{P}(\rho)$ is the final state condition, and $R \subseteq \mathcal{P}(\sigma \times \rho) \times (\mathcal{F}(\sigma \times \rho \rightarrow \gamma))^* \times \mathcal{F}(\sigma \times \rho \rightarrow \rho)$ is a finite set of rules \square

We write $A^{\sigma/\gamma:\rho}$ to indicate the input/output element type σ/γ and the register type ρ of an ST A . When we write $A^{\sigma/\gamma}$, we assume ρ to be implicit. A rule $(\varphi, u, r) \in R_A$ denotes the following set of concrete transitions:

$$\llbracket (\varphi, u, r) \rrbracket \stackrel{\text{def}}{=} \{q \xrightarrow{a/\llbracket u \rrbracket(a,q)} \llbracket r \rrbracket(a, q) \mid (a, q) \in \llbracket \varphi \rrbracket\}$$

Although the formal definition omits finite states, it is often useful to explicitly include a separate finite state component, we do this in the examples below. Moreover, it is technically convenient to extend final states with *final outputs* by extending ϑ to be a finite set of *final output rules* as a subset of $\mathcal{P}(\rho) \times (\mathcal{F}(\rho \rightarrow \gamma))^*$ such that if $(\psi, v) \in \vartheta$ and $q \in \llbracket \psi \rrbracket$ then a final output produced from q is $\llbracket v \rrbracket(q)$ (recall that $\llbracket v \rrbracket$ denotes the function $\lambda q. \llbracket v[0] \rrbracket(q), \dots, \llbracket v[v-1] \rrbracket(q) \rrbracket$, denoted $q \xrightarrow{\llbracket v \rrbracket(q)} A$). Note that, when all final outputs are ϵ then ϑ is equivalent to being a ρ -predicate as in Definition 8, i.e., $q \in \llbracket \vartheta_A \rrbracket$ then means that $q \xrightarrow{\epsilon} A$. Final outputs correspond to a restricted form of input-epsilon rules as used in classical finite transducers.

The reachability relation $p \xrightarrow{\mathbf{u}/\mathbf{v}}_A q$ for $\mathbf{u} \in \Sigma^*$, $\mathbf{v} \in \Gamma^*$, and $p, q \in \mathcal{U}^\rho$ is defined analogously to SFTs. The definition of \mathcal{T}_A is lifted similarly, for $\mathbf{u} \in \Sigma^*$:

$$\mathcal{T}_A(\mathbf{u}) \stackrel{\text{def}}{=} \{\mathbf{v} \cdot \mathbf{w} \mid \exists q (q_A \xrightarrow{\mathbf{u}/\mathbf{v}}_A q \wedge q \xrightarrow{\mathbf{w}}_A)\}$$

Example 7: Consider the C# code in Example 3. There is a direct mapping of the code to an ST A that uses the compound register $\langle q, c \rangle$. The initial state of A is $\langle 0, 0 \rangle$, the final state condition is \mathbf{t} and the rules are:

$$\begin{aligned} (q=0 \wedge x=\langle ' < ' , \epsilon, \langle 1, x \rangle), & & (q=0 \wedge x \neq \langle ' < ' , \epsilon, \langle 0, x \rangle), \\ (q=1 \wedge x=\langle ' < ' , \epsilon, \langle 1, x \rangle), & & (q=1 \wedge x \neq \langle ' < ' , \epsilon, \langle 2, x \rangle), \\ (q=2 \wedge x=\langle ' > ' , [\langle ' < ' , c, \langle ' > ' \rangle], \langle 0, x \rangle), & & (q=2 \wedge x \neq \langle ' > ' , \epsilon, \langle 0, x \rangle) \end{aligned}$$

The register update $\langle r_q, r_c \rangle$ of a rule corresponds to the assignments $q := r_q$ and $c := r_c$. Since all assignments to c have the form $c := x$, c corresponds to the *previous* input character. A can be automatically transformed to the equivalent *GetTags* SFT; the register c is eliminated by using a new state and nondeterminism. \square

One can effectively construct a well-founded axiomatic theory $Th(A)$ of an ST $A^{\sigma/\gamma}$ over a background of *lists*, similarly to symbolic automata in [46]. $Th(A)$ defines a symbol T_A that provides a sound and complete axiomatization of \mathcal{T}_A , i.e., for any model $\mathfrak{A} \models_{\mathcal{U}} Th(A)$, $T_A^{\mathfrak{A}} = \mathcal{T}_A$.

Moreover, $Th(A)$ can be directly asserted as an *auxiliary theory* of any state-of-the-art SMT solver that supports lists. By deploying $Th(A)$ in this way, we obtain an *integrated decision procedure* for satisfiability and model generation for quantifier free formulas that may arbitrarily combine formulas over the background \mathcal{U} with *transduction atoms* $T_A(u, v)$ where $u : \mathbb{L}(\sigma)$ and $v : \mathbb{L}(\gamma)$ are arbitrary list terms. A direct application, outlined in Figure 5, is a semi-decision

```

Witness1disequality( $A^{\sigma/\gamma}, B^{\sigma/\gamma}$ )  $\stackrel{\text{def}}{=}
1 \text{ assert } Th(A) \cup Th(B);
2 (u, v, w) := (nil, NewVariable(\mathbb{L}(\gamma)), NewVariable(\mathbb{L}(\gamma)));
3 \text{ while } \mathbf{t}
4 \text{ if } \exists \mathfrak{A} (\mathfrak{A} \models_{\mathcal{U}} T_A(u, v) \wedge T_B(u, w) \wedge v \neq w) \text{ return } (u^{\mathfrak{A}}, v^{\mathfrak{A}}, w^{\mathfrak{A}});
5 \text{ else } u := cons(NewVariable(\sigma), u);$ 
```

Figure 5: Given STs $A \not\equiv B$, generates a witness (u, v, w) such that $v \in \mathcal{T}_A(u)$, $w \in \mathcal{T}_B(u)$, and $v \neq w$.

procedure for *1-disequality* of STs, where the auxiliary theories are asserted to the solver in line 1, and successively longer input-lists are used to invoke the solver to decide 1-disequality of the instance in line 4. The semi-decision procedure computes a shortest-input witness of 1-disequality.

4. CASE STUDIES

We present four case studies for applications of SFTs. While the first case study focuses on sophisticated string manipulation, going beyond previous approaches such as BEK [4], we want to emphasize that the utility of SFTs goes well beyond reasoning about string sanitizer processing. Figure 2 summarizes the essential features of each case study.

4.1 Representing HTMLDecode

To prevent injection attacks such as cross-site scripting (XSS) and SQL injection, Web applications employ *sanitizers*, which are string manipulation routines that remove or encode dangerous input characters. Many applications include their own sanitizer implementations. Recent work by Hooimeijer et al. [4] examines several such sanitizers, demonstrating that a subset of popular sanitizers can be modeled using transducers. Furthermore, they show that safety properties of Web sanitizers can be checked using transducer analyses.

We focus on the sanitizer `HTMLDecode` to evaluate the practical utility of the ST representation. Figure 6 outlines a real-world implementation, taken from the OWASP library. `HTMLDecode` transforms HTML *entities* back to the symbol they represent. Entities can be named (e.g., `<`; maps to `<`), or numeric in decimal or hexadecimal representation (e.g., decimal entity `0`; maps to symbol `0`). For simplicity, we will restrict our attention to decimal entities.

Intuitively, `HTMLDecode` is difficult to cast as a transducer because it requires *lookahead*: a single output symbol may depend on a specific sequence of several characters. The full Unicode set consists of more than one million symbols. To decode a decimal entity, therefore, we need to inspect up to six digits. While that is possible using either SFTs or traditional transducers, it requires a large state space.

```

public String HTMLDecode( String input ) {
  StringBuffer sb = new StringBuffer();
  PushbackString pbs = new PushbackString( input );
  while ( pbs.hasNext() ) {
    Character c = decodeCharacter( pbs );
    if ( c != null ) { sb.append( c ); }
    else { sb.append( pbs.next() ); } }
  return sb.toString();
}

public Character decodeCharacter( PushbackString input ) {
  input.mark();
  Character first = input.next();
  if ( first == null ) { input.reset(); return null; }
  if ( first.charValue() != '&' ) {
    input.reset(); return null; }
  Character second = input.next(); if ( second == null ) {
    input.reset(); return null; }
  if ( second.charValue() == '#' ) {
    Character c = getNumericEntity( input );
    if ( c != null ) return c; }
  else if ( Character.isLetter( second.charValue() ) ) {
    input.pushback( second );
    Character c = getNamedEntity( input );
    if ( c != null ) return c; }
  input.reset();
  return null;
}

private Character getNumericEntity( PushbackString input ) {
  ...
  return parseNumber( input );
}

private Character parseNumber( PushbackString input ) {
  StringBuffer sb = new StringBuffer();
  while( input.hasNext() ) {
    Character c = input.peek();
    if ( Character.isDigit( c.charValue() ) ) {
      sb.append( c );
      input.next(); }
    else if ( c.charValue() == ';' ) {
      input.next();
      break; }
    else break; }
  try {
    int i = Integer.parseInt(sb.toString());
    return new Character( (char)i ); }
  catch( NumberFormatException e ) return null;
}

```

Figure 6: Excerpt `HTMLDecode` in Java (from OWASP 1.4.0). The code shown converts named entities (e.g., `<` to `<`) and numeric entities (e.g., `4`; to `4`). The numeric entity conversion is difficult to model efficiently using previous approaches.

In contrast, the corresponding ST is quite succinct. Figure 7 shows a ST $Decode_{\mathbb{Z}/\mathbb{Z}; Q \times (\mathbb{Z} \times \mathbb{Z})}$, that uses two registers to handle numeric entities with exactly two digits. The compound register is $\langle q, \langle y, z \rangle \rangle$. We illustrate explicitly the finite state component (that is the value of q). Final outputs, unless they are ϵ , are shown by labels on outgoing \rightarrow arcs from final states. Characters correspond to their Unicode code points, e.g., `'0'` = 48. The actual decoding happens in the rule from state q_4 to q_0 with guard $x = \text{';'}$, where the output $10 \cdot (y - 48) + z - 48$ corresponds to invocation of `parseInt` in Figure 6. The states q_i roughly correspond to the control flow of the code in Figure 6.

Evaluation: We compare the ST representation to the equivalent SFTs, in terms of size and analysis speed. Let $DecodeST_n$ denote an ST that models `HTMLDecode` for entities of the form `&#[0-9]{1,n}`; (i.e., up to n decimals, inclusive). For each $i \in [2; 6]$, we compute an SFT that is equivalent to $DecodeST_i$ by concretizing the possible register values at each state. Figure 8(a) shows the number of

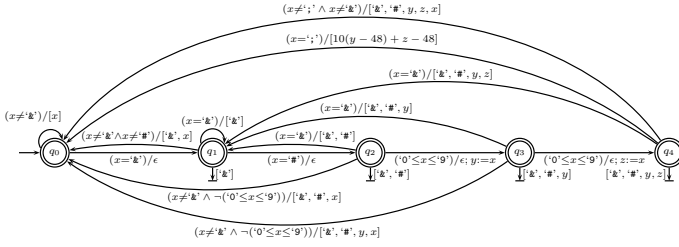


Figure 7: ST representation of the HTMLdecode code in Figure 6, restricted to decimal numeric entities of the form `&#[0-9][0-9]`; . This ST uses two registers to remember one digit each, and uses integer-linear arithmetic to compute the corresponding code point.

both states and edges on the y -axis, for both representations; the x -axis denotes the number of digits modeled. The most prominent take-away is that the ST representation is drastically smaller. For example, for the 6-digit encoding, the SFT encoding has over $10,000\times$ as many states, and $135,000\times$ as many edges, as the equivalent ST encoding.

We consider the speed of two algorithms: composition and equivalence checking. The experimental task is to find a witness (i.e., an input string) w that demonstrates that $\text{DecodeST}_i(w) \neq \text{DecodeST}_i(\text{DecodeST}_i(w))$, for some number of self-compositions. We consider three different algorithms for performing this task: ST.E uses an in-memory representation and conducts an *eager* search (computing entire transducers); ST.L uses a *lazy* approach by encoding the entire ST into the underlying SMT solver; and SFT represents an eager SFT-based approach.

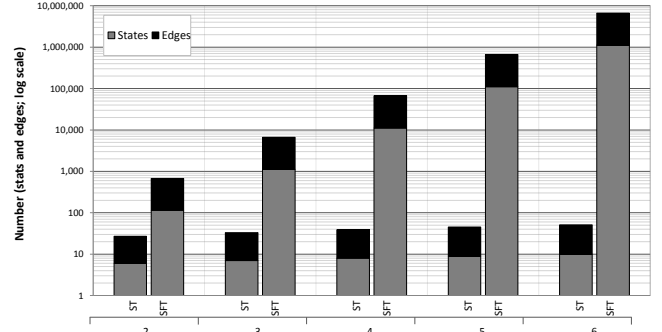
Figure 8(b) shows the speed results. Compositions represents the number of times we compose each class of transducer with itself. Composition refers to the time taken to perform the composition; for the lazy ST.L this time is negligible (since the composition is simply asserted to the underlying solver). IdempotenceChecking refers to the time taken to find the actual witness, after the composition. Each column represents a single experimental run; we employed a 2-hour timeout per run. The label OM marks runs that ran out of memory, while TO marks cases that hit the 2-hour timeout.

These results show that STs, in particular using the lazy representation, are significantly more scaleable for this task than SFTs. The SFT representation either exhausts memory or passes the timeout in the majority of cases. The eager ST representation outperforms the lazy representation only for the smallest two testcases. For larger runs (i.e., more compositions and more digits), the lazy SFT representation scales much more reliably, ranging from 1 to 20 seconds over the 2-composition range (compared to several minutes for the eager representation).

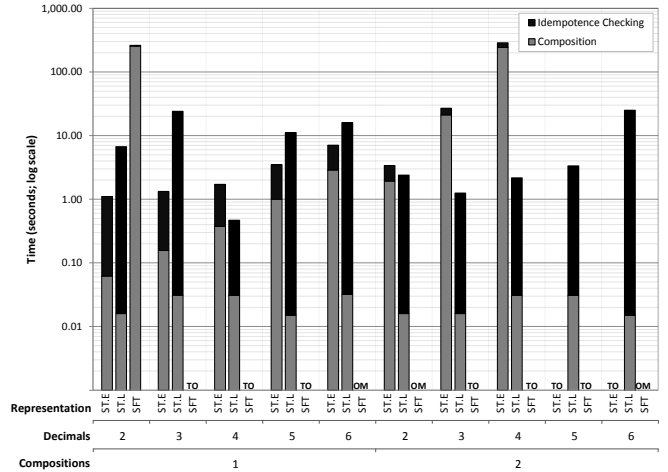
4.2 Malware Fingerprinting Code

Millions of web pages today contain malicious JavaScript that attempts to take over a victim’s web browser. An active research literature has proposed static and dynamic methods for detecting these attacks [40, 13, 11, 9]. A key finding of this work is that malware authors use *fingerprinting* techniques [35, 16] to decide which malware to deliver to the victim user.

Figure 9 shows an example of client-side browser fin-



(a) Initial transducer sizes.



(b) Running times.

Figure 8: HTMLDecode results. The task is to prove that HTMLDecode does not commute with itself, and to provide a witness that demonstrates this for a given number of Compositions. We evaluate three representations: ST.E (eager ST composition), ST.L (lazy ST composition using Z3), and SFT (SFT composition). We consider five distinct models (indicated by Decimals, based on how many digits the ST or SFT can handle).

gerprinting.¹ The code iterates over the list of plugins installed in the browser and queries their version numbers. In some cases, version numbers are padded by optionally adding leading 0s to them. Finally, variables `quicktime_plugin` and `adobe_plugin` are combined to produce the final `fingerprint` value. This fingerprint value is then used to select a specific attack to run against the user.

Figure 12 list several concrete fingerprint values from real browser setups. Note that QuickTime 7.6.6 has at least one known vulnerability, and may thus be of special interest to an attacker.

We consider a scenario in which we have acquired these fingerprints (e.g., through network sniffing), and want to find out the corresponding plugin names. At a higher level, the question is: “Can we find out interesting properties by computing string-related pre-conditions based on a postcondition?”

Our techniques can answer this question in the affirmative by modeling the code of Figure 9 using multiple SFTs.

¹This code is simplified for illustrative purposes: the original considers more plugin types, including Flash, etc.


```

var quicktime_plugin = "0", adobe_plugin = "00";

for(var i = 0; i < navigator.plugins.length; i++)
{
  var plugin_name = navigator.plugins[i].name;
  if (quicktime_plugin == 0 &&
      plugin_name.indexOf("QuickTime") != -1)
  {
    var helper = parseInt(plugin_name.replace(/\/D/g, ""));
    if (helper > 0) // not base 16
      quicktime_plugin = helper.toString(10)
  }
  if (adobe_plugin == "00" &&
      plugin_name.indexOf("Adobe Acrobat") != -1)
  {
    plugin_name = navigator.plugins[i].description;
    if(plugin_name.indexOf(" 5") != -1) adobe_plugin = "05";
    else if(plugin_name.indexOf(" 6") != -1) adobe_plugin = "06";
    else if(plugin_name.indexOf(" 7") != -1) adobe_plugin = "07";
    else adobe_plugin = "01"
  } else {
    // flash, java...
  }
}

while(quicktime_plugin.length < 8)
  quicktime_plugin = "0" + quicktime_plugin;

var fingerprint = "Q" + quicktime_plugin + "8" + adobe_plugin;
// ...
fetch_exploit(fingerprint);

```

Figure 9: Browser and plugin fingerprinting code found in JavaScript malware.

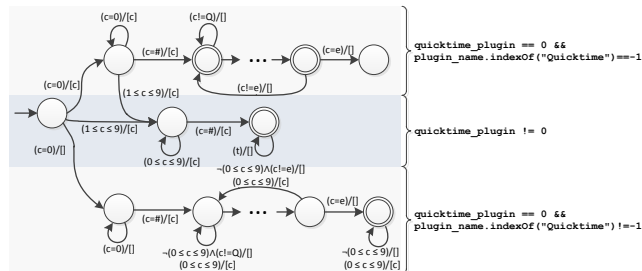


Figure 10: SFT QuicktimeSplitter with corresponding path predicates for the fingerprinting code.

The key idea is conditional assignment translates into non-deterministic case splits inside the SFT. At a high level, each transducer corresponds to a split or a merge in the control flow of the fingerprinting code, relative to a single variable of interest. This is illustrated in Figure 10, which shows the transducer `QuicktimeSplitter` together with the path predicates modeled by each of its three main branches. The transducer reads both `quicktime_plugin` and `plugin_name`, separated by a special `#` symbol. Its output is guaranteed to start with `#` if and only if the branch `if(quicktime_plugin == 0 && ...)` was taken.

For the sake of brevity, we do not display the remaining SFTs. Figure 11 lists the full set of SFTs and their statistics. `QuicktimeMerger` takes the output of `QuicktimeSplitter` and models the control flow join at the end of the first `if` statement. `QuickTimePadder` models the final `while` loop in Figure 9. The manipulation of variable `adobe_plugin` is analogous.

Evaluation: We compute the pre-image of the composition transducers discussed above as follows. For each fingerprint w , we construct an SFA that accepts $\{w\}$ and corresponds to the postcondition `fingerprint == w`. The pre-

SFT	States	Edges
Quicktime (variable quicktime_plugin)		
QuicktimeSplitter	25	60
QuicktimeMerger	6	9
QuicktimePadder	37	37
Composed	534	1,425
Adobe (variable adobe_plugin)		
AdobeSplitter	36	81
AdobeMerger	21	40
Composed	203	797

Figure 11: SFTs used for the malware fingerprinting example. Statistics for the Quicktime and Adobe components are shown. The composition SFTs take approximately one second to compute.

Browser/plugin combination	Fingerprint
FF: Acrobat 9.4.5.236; no quicktime	Q00000000801
FF: Acrobat 9.4.5.236; Quicktime 7.6.9	Q00000769801
IE: no plugins of interest installed	Q00000000800
FF: Acrobat 9.4.5.236; Quicktime 7.5.5	Q00000755801
FF: Acrobat 9.4.5.236; Quicktime 7.6.6	Q00000766801

Figure 12: Browser fingerprints. Using an SFT model, computing input values for `quicktime_plugin` and `adobe_plugin` takes less than one second per fingerprint.

images of `QuicktimeComposed` and `AdobeComp` correspond to preconditions for a single iteration of the `for` loop in Figure 9. For example, for $w = Q00000769801$, we find a precondition that relates values of `quicktime_plugin` to values of `plugin_name`, as follows: (1) `quicktime_plugin` already had value 769, possibly padded with up to five zeroes, or (2) `quicktime_plugin` consisted entirely of zeroes and `plugin_name` contained the substring `Quicktime` together with digits 7, 6, and 9 in that relative order. Condition (1) represents the case where `quicktime_plugin` had a previously-assigned version number, while condition (2) represents the case in which a version number was extracted inside the `for` loop. For all real fingerprints we tried, our analysis took less than one second per fingerprint.

Next, we evaluate whether inverse image generation, as used above, scales to relatively large output values. Unlike most previous string constraint solvers [29, 41, 26], SFT-

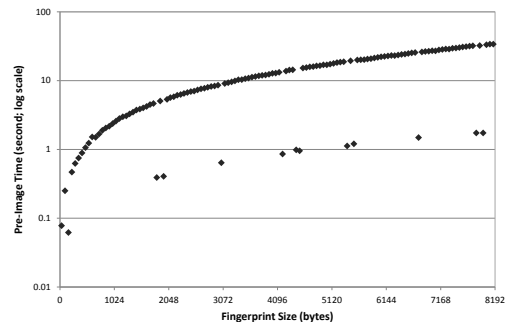


Figure 13: Inverse image generation time in seconds for fingerprint outputs up to 8192 bytes. The outputs were randomly generated from the language $Q[0-9]\{n\}801$ over 16 bit characters.

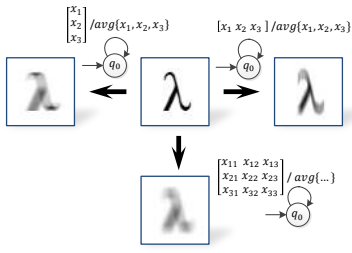


Figure 14: Blurring transformation illustrated.

based analysis does not impose length bounds on the strings under consideration. This is only beneficial if the approach actually scales to large strings. We show the approach *does* scale by generating random fingerprints of the form $Q[0-9]\{n\}801$, and measuring the time it takes to compute the inverse images for both the QuickTime and Adobe variables. Figure 13 shows encouraging results: in general, our approach takes less than half a minute to generate pre-images for up to *eight kilobytes* worth of output. In contrast, the Hampi solver was limited to finding pre-images up to fifty bytes worth of output.

Our malware case study demonstrates several important points. First, SFTs are well-suited for describing code by making use of non-determinism. The transducers needed can be large (on the order of hundreds of states and edges), but we can construct them from much smaller transducers (tens of states and edges) through composition. The pre-image computation reveals interesting relations among mutually dependent variables. Finally, the pre-image computation is efficient: it can generate valid string inputs for outputs that measure several kilobytes in size, while we are unaware of any previous string constraint solver that can handle this order of magnitude.

Takeaways: From our first two case studies, we have the following key takeaways

- STs can be radically more succinct in representation than SFTs: we saw in our HTMLDecode example that our ST representation had 10,000 times fewer states and 150,000 times fewer edges than our SFT representation.
- Lazy ST encoding scales best for our HTMLDecode example, taking between 1 to 20 seconds for six characters and two compositions. Eager ST encoding is slower, and eager SFT encoding times out above two characters.
- SFTs can accurately model real examples of malicious Javascript fingerprinting code. Our analysis requires less than one second to recover plug-in versions from real examples of fingerprints generated by malicious code found “in the wild.”
- Previous work in string constraint solving has focused almost exclusively on constraints that require fewer than 50 bytes; for example, the majority of Hampi experiments were conducted with length bounds of 15 bytes or fewer [29]. In contrast, our techniques can synthesize pre-images in excess of 8,000 bytes in roughly a minute.

4.3 Image Blurring

To illustrate the generality of SFTs, we look at image transformations. A clear advantage of representing image transformations in the form of transducers is the ability to do composition on transducers it gives us. In fact, image editors such as Google Picasa represent image contents as the original image as well as a series of image transformations, such as blurring, sharpening, black and white conversion, contrast enhancement, and the like.

In many cases, of course, editing a large-scale image that includes millions of 32-bit pixels before high-quality printing might involve a dozen of such transformations. Applying them one after another, in a sequence is often too time-consuming to be practical. A better alternative consists of composing the transformations together and applying them to the input image only a single time.

We focus on *image blurring*, which is a prototypical “textbook” image transformation [39]. Figure 14 illustrates the two transducers for *horizontal* and *vertical* blurring of an image.

Measuring privacy via entropy: A fascinating feature of our analysis is that we can estimate a *privacy* metric for image blurring using our techniques. Our starting point is the observation that if an image has a *unique* pre-image given blurring, then the blurring does not hide the original image at all. Just consider a black square: no matter how many times we might attempt to blur it, the image will remain unchanged. On the other hand, after blurring a face, there may be multiple original images that yield the same blurred face.

Put another way, a blurred face image defines a set of potential candidate original images. If we assume that all candidate original images are equally likely, then we can define the *entropy* H of the original face after a blurring transformation β as follows:

$$H = -\log(|\beta(\beta^{-1})|)$$

in other words, we use the reverse mapping β^{-1} given to us by the inverse transducer, and take the negated logarithm of the size of its preimage. For the entirely black image example, $H = 0$ because there is only one element in the preimage. To increase privacy, our goal is to maximize the entropy. Note that given for images of a given rectangular size, we can always exhaustively check if two images result in the same image after blurring.

Our techniques allow us to write down a SMT formula where the number of solutions is equal to the number of preimages of β on a specific image. This is the first connection to our knowledge between SMT techniques and probabilistic definitions of privacy. Of course computing the exact number of solutions is $\#P$ -complete, but we can employ approximation techniques to estimate this quantity. Then we can programmatically compare different methods of blurring by the entropy induced. We leave exploration of the impact of different approximation techniques as future work.

4.4 Location Privacy

GPS sensors located in most mobile devices constantly track our location [31, 2]. While the popularity of applications such as Foursquares, Gowalla, and Facebook check-ins show user demand for sharing this information, this also raises privacy concerns.

As a reaction to privacy concerns, Google’s Latitude location sharing service started allowing users to only release the *city* they are in, not their precise location, so that a trace of

a user’s day might look like

Menlo Park, CA; Palo Alto, CA; Los Gatos, CA;
Mountain View, CA; Los Gatos, CA.

Clearly, given the sizes of these cities, tracking the user precisely might present some difficulty. To generalize, one approach that has emerged is context sensitive choice of granularity; intuitively, if I am located in a densely-populated location such as midtown Manhattan, block-level location is fine to reveal. If I am in a sparsely-populated area such as the Death Valley, we should only reveal a coarse location approximation. This can be encoded with a transducer that works on a stream of recorded location measurements. To summarize, given GPS coordinates (latitude/longitude):

1. Use a lookup list of world cities and their latitude/longitude values and nearest point calculation to compute the nearest city to the current location;
2. Determine the city population via a lookup table;
3. Map the population to a high or low density area (H or L).
4. Based on the last five GPS readings, enter a high- or low-density output state. Depending on that approximate the (latitude,longitude) pair with different precision.

This can be captured by a transducer with different output actions depending on the current state. For instance, for high-density areas, we can drop GPS location seconds, and for low-density areas we can drop GPS location minutes. Krumm et al consider additional trace obfuscation techniques such as adding noise or quantizing traces, which can also be represented using our SFT framework [8].

5. RELATED WORK

General equivalence of finite state transducers is undecidable [22], and already so for very restricted fragments [27]. Equivalence of decidability of single-valued GSMs was shown in [42], and extended to the finite-valued case (there exists k such that, for all v , $|\mathcal{T}_A(v)| \leq k$) in [12, 49]. The decidability of equivalence of the finite-valued case does not follow from the single-valued case. Corresponding decidability result of equivalence of finite-valued SFTs is shown in [6]. Unlike for the single-valued case that has a practical algorithm (Figure 3), the finite-valued case is substantially harder, the 1-equality algorithm does not generalize to this case because the satisfiability checks cannot be made locally: Lemma 2 does not imply violation of partial-equivalence in the finite-valued case.

In recent years there has been considerable interest in automata over infinite languages [43], starting with the work on *finite memory automata* [28], also called *register automata*. Finite words over an infinite alphabet are often called *data words* in the literature. Other automata models over data words are *pebble automata* [36] and *data automata* [7]. Several characterizations of logics with respect to different models of data word automata are studied in [5]. This line of work focuses on fundamental questions about definability, decidability, complexity, and expressiveness on classes of automata on one hand and fragments of logic on the other hand. A different line of work on automata with infinite alphabets introduces *lattice automata* [21] that are finite state automata whose transitions are labeled by elements of an atomic lattice with motivation coming from verification of symbolic communicating machines.

Streaming transducers [1] provide another recent symbolic extension of finite transducers where the label theories are restricted to be total orders, in order to maintain decidability of equivalence, e.g., full linear arithmetic is not allowed.

Finite state automata with arbitrary predicates over labels, called *predicate-augmented finite state recognizers*, or *symbolic finite automata* (SFAs) in the current paper, were first studied in the context of natural language processing [37]. While the work [37] views symbolic automata as a “fairly trivial” extension, the fundamental algorithmic questions are far from trivial. For example, it is shown in [24] that symbolic complementation by a combinatorial optimization problem called *minterm generation* leads to significant speedups compared to state-of-the-art automata algorithm implementations. The work in [37] introduces a different symbolic extension to finite state transducers called *predicate-augmented finite state transducers*. This extension is not expressive enough for describing SFTs. Besides identities, it is not possible to establish functional dependencies from input to output that are needed for example to encode transformations such as `HtmlEncode`.

SFTs are used in the BEK project [23] and use the SMT solver Z3 [15] for solving label constraints that arise during composition and equivalence checking algorithms, as well as for witness search by model generation using auxiliary SFT axioms. Finite state transducers have been used for dynamic and static analysis to validate sanitization functions in web applications in [3], by an over-approximation of the strings accepted by the sanitizer using static analysis of existing PHP code. Other security analysis of PHP code, e.g., SQL injection attacks, use string analyzers to obtain over-approximations (in form of context free grammars) of the HTML output by a server [34, 48]. Yu et al. show how multiple automata can be composed to model looping code [50].

Our work is complementary to previous efforts in using SMT solvers to solve problems related to list transformations. HAMPI [29] and Kaluza [41] extend the STP solver to handle equations over strings and equations with multiple variables. The work in [25] shows how to solve subset constraints on regular languages. In contrast, we show how to combine any of these solvers with SFTs whose edges can take symbolic values in the theories understood by the solver.

Top-down tree transducers [20] provide another extension of finite state transducers: a finite state transducer is a top-down tree transducer over a *monadic* ranked alphabet. Similar to finite state transducers, decidability of equivalence of top-down tree transducers is known for the single-valued case [17, 19], including a specialized method for the *deterministic* case [10], and also for the finite-valued case [44]. Several non-symbolic extensions of top-down tree transducers have been studied, e.g., [20, 33, 18, 32, 30, 38]. Symbolic top-down tree transducers are studied in [45] where partial equivalence is shown to be decidable for the linear single-valued case.

6. CONCLUSION

We introduced a symbolic extension of the theory of classical finite transducers, where transitions are represented by terms modulo a given background theory. Our approach enables a range of analyses in combination with state-of-the-art constraint solving techniques. The core algorithms we presented are composition and equivalence checking of single-valued symbolic finite transducers, and we showed how to decide whether arbitrary symbolic transducers have the single-valuedness property. We demonstrated how our

work directly applies to analysis of web string sanitizers, malware detection, image manipulation, and location privacy, and we expect more applications to follow. Our techniques can synthesize string pre-images in excess of 8,000 bytes in roughly a minute, our ST representation had 10,000 times fewer states than previous approaches, and we found lazy ST encoding for our HTMLDecode example took at most 20 seconds even in the most extreme cases. These algorithms make it possible to work with symbolic representations of transducers, just as traditionally done with finite state transducers, as first class citizens in designing new analyses and program transformation techniques by leveraging the continuous advances and improvements in constraint solvers and satisfiability modulo theories solvers.

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