SNAP: Stateful Network-Wide Abstractions for Packet Processing

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ABSTRACT

Early programming languages for software-defined networking (SDN) were built on top of the simple match-action paradigm offered by OpenFlow 1.0. However, emerging hardware and software switches offer much more sophisticated support for persistent state in the data plane, without involving a central controller. Nevertheless, managing stateful, distributed systems efficiently and correctly is known to be one of the most challenging programming problems. To simplify this new SDN problem, we introduce SNAP.

SNAP offers a simpler "centralized" stateful programming model, by allowing programmers to develop programs on top of *one* big switch rather than *many*. These programs may contain reads and writes to global, persistent arrays, and as a result, programmers can implement a broad range of applications, from stateful firewalls to fine-grained traffic monitoring. The SNAP compiler relieves programmers of having to worry about how to distribute, place, and optimize access to these stateful arrays by doing it all for them. More specifically, the compiler discovers read/write dependencies between arrays and translates one-big-switch programs into an efficient internal representation based on a novel variant of binary decision diagrams. This internal representation is used to construct a mixed-integer linear program, which jointly optimizes the placement of state and the routing of traffic across the underlying physical topology. We have implemented a prototype compiler and applied it to about 20 SNAP programs over various topologies to demonstrate our techniques' scalability.

CCS Concepts

•**Networks** → **Programming interfaces; Programmable networks;** *Network control algorithms; In-network processing; Network management;*

Keywords

SNAP, Network Programming Language, Stateful Packet Processing, One Big Switch, Software Defined Networks, Optimization

1. INTRODUCTION

The first generation of programming languages for software-defined networks (SDNs) [14, 10, 44, 20, 17] was built on top of OpenFlow 1.0, which offered simple match-action processing of packets. As a result, these systems were partitioned into (1) a stateless packet-processing part that could be analyzed statically, compiled, and installed on OpenFlow switches, and (2) a general stateful component that ran on the controller.

This "two-tiered" programming model can support any network functionality by running the stateful portions of the program on the controller and modifying the stateless packet-processing rules accordingly. However, simple stateful programs, such as detecting SYN floods or DNS amplification attacks, cannot be implemented *efficiently* because packets must go back-and-forth to the controller, incurring significant delay. Thus, in practice, stateful controller programs are limited to those that do not require *per-packet* stateful processing.

Today, however, SDN technology has advanced considerably: there is a raft of new proposals for switch interfaces that *expose persistent state on the data plane*, including those in P4 [6], OpenState [4], POF [38], Domino [33], and Open vSwitch [25]. Stateful programmable data planes enable us to *offload* programs that require *per-packet* stateful processing onto switches, subsuming a variety of functionality normally relegated to middleboxes. However, the mere existence of these stateful mechanisms does not make networks of these devices easy to program. In fact, programming distributed collections of stateful devices is typically one of the most difficult kinds of programming problems. We need new languages and abstractions to

help us manage the complexity and optimize resource utilization effectively.

For these reasons, we have developed SNAP, a new language that allows programmers to mix primitive stateful operations with pure packet processing. However, rather than ask programmers to program a large, distributed collection of independent, stateful devices manually, we provide the abstraction that the network is *one* big switch (OBS). Programmers can allocate persistent arrays on that OBS, and do not have to worry about where or how such arrays are stored in the physical network. Such arrays can be indexed by fields in incoming packets and modified over time as network conditions change. Moreover, if multiple arrays must be updated simultaneously, we provide a form of *network transaction* to ensure such updates occur atomically. As a result, it is easy to write SNAP programs that learn about the network environment and record its state, store per-flow information or statistics, or implement a variety of stateful mechanisms.

While it simplifies programming, the OBS model, together with the stateful primitives, generates implementation challenges. In particular, multiple flows may depend upon the same state. To process these flows correctly and efficiently, the compiler must simultaneously determine which flows depend upon which components, how to route those flows, and where to place the components. Hence, to map OBS programs to concrete topologies, the SNAP compiler discovers read-write dependencies between statements. It then translates the program into an xFDD, a variant of forwarding decision diagrams (FDDs) [35] extended to incorporate stateful operations. Next, the compiler generates a system of integer-linear equations that jointly optimizes array placement and traffic routing. Finally, the compiler generates the switch-level configurations from the xFDD and the optimization results. We assume that the switches chosen for array placement support persistent programmable state; other switches can still play a role in routing flows efficiently through the state variables. Our main contributions are:

- A stateful and compositional SDN programming language with persistent global arrays, a one-big-switch programming model, and network transactions. (See §2 for an overview and §3 for more technical details.)

- Algorithms for compiling SNAP programs into low-level switch mechanisms (§4): (i) an algorithm for compiling SNAP programs into an intermediate representation that detects program errors, such as race conditions introduced by parallel access to stateful components, using our extended forwarding decision diagrams (xFDD) and (ii) an algorithm to generate a mixed integer-linear program, based on the xFDD, which jointly decides array placement and routing while minimizing network congestion and satisfying the constraints necessary for network transactions.

DNS-tunnel-detect

```
1   if dstip = 10.0.6.0/24 & srcport = 53 then
2     orphan[dstip][dns.rdata] <- True;
3     susp-client[dstip]++;
4     if susp-client[dstip] = threshold then
5       blacklist[dstip] <- True
6     else id
7   else
8     if srcip = 10.0.6.0/24 & orphan[srcip][dstip]
9     then orphan[srcip][dstip] <- False;
10         susp-client[srcip]--
11     else id
```

Figure 1: SNAP implementation of `DNS-tunnel-detect`.

- An implementation and evaluation of our language and compiler using about 20 applications. (§5, §6).

We discuss various data-plane implementations for SNAP, how SNAP relates to middleboxes, and possible extensions in §7, discuss related work in §8, and conclude in §9.

2. SNAP SYSTEM OVERVIEW

This section overviews the key concepts in our language and compilation process using example programs.

2.1 Writing SNAP Programs

DNS tunnel detection. The DNS protocol is designed to resolve information about domain names. Since it is not intended for general data transfer, DNS often draws less attention in terms of security monitoring than other protocols, and is used by attackers to bypass security policies and leak information. Detecting DNS tunnels is one of many real-world scenarios that require state to track the properties of network flows [5]. The following steps can be used to detect DNS tunneling [5]:

1. For each client, keep track of the IP addresses resolved by DNS responses.

2. For each DNS response, increment a counter. This counter tracks the number of resolved IP addresses that a client does not use.

3. When a client sends a packet to a resolved IP address, decrement the counter for the client.

4. Report tunneling for clients that exceed a threshold for resolved, but unused IP addresses.

Figure 1 shows a SNAP implementation of the above steps that detects DNS tunnels to/from the CS department subnet 10.0.6.0/24 (see Figure 2). Intuitively, a SNAP program can be thought of as a function that takes in a packet plus the current state of the network and produces a set of transformed packets as well as updated state. The incoming packet is read and written by referring to its fields (such as `dstip` and `dns.rdata`). The "state" of the network is read and written by referring to user-defined, array-based, global variables

(such as `orphan` or `susp-client`). Before explaining the program in detail, note that it does not refer to specific network device(s) on which it is implemented. SNAP programs are expressed as if the network was *one-big-switch* (OBS) connecting edge ports directly to each other. Our compiler automatically distributes the program across network devices, freeing programmers from such details and making SNAP programs portable across topologies.

The `DNS-tunnel-detect` program examines two kinds of packets: incoming DNS responses (which may lead to possible DNS tunnels) and outgoing packets to resolved IP addresses. Line 1 checks whether the input packet is a DNS response to the CS department. The condition in the `if` statement is an example of a simple *test*. Such tests can involve any boolean combination of packet fields.[1] If the test succeeds, the packet could potentially belong to a DNS tunnel, and will go through the detection steps (Lines 2–6). Lines 2–6 use three global variables to keep track of DNS queries. Each variable is a mapping between keys and values, persistent across multiple packets. The `orphan` variable, for example, maps each pair of IP addresses to a boolean value. If `orphan[c][s]` is `True` then `c` has received a DNS response for IP address `s`. The variable `susp-client` maps the client's IP to the number of DNS responses it has received but not accessed yet. If the packet is not a DNS response, a different test is performed, which includes a stateful test over `orphan` (Lines 8). If the test succeeds, the program updates `orphan[srcip][dstip]` to `False` and decrements `susp-client[srcip]` (Lines 10–11). This step changes the global state and thus, affects the processing of future packets. Otherwise, the packet is left unmodified — `id` (Line 12) is a no-op.

Routing. `DNS-tunnel-detect` cannot stand on its own—it does not explain where to forward packets. In SNAP, we can easily *compose* it with a forwarding policy. Suppose our target network is the simplified campus topology depicted in Figure 2. Here, I_1 and I_2 are connections to the Internet, and D_1–D_4 represent edge switches in the departments, with D_4 connected to the CS building. C_1–C_6 are core routers connecting the edges. External ports (marked in red) are numbered 1–6 and IP subnet `10.0.i.0/24` is attached to port `i`. The `assign-egress` program assigns outports to packets based on their destination IP address:

```
assign-egress = if dstip = 10.0.1.0/24
then outport <- 1
else if dstip = 10.0.2.0/24 then outport <- 2
else ...
else if dstip = 10.0.6.0/24 then outport <- 6
else drop
```

Note that the policy is independent of the internal

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[1] The design of the language is unaffected by the chosen set of fields. For the purposes of this paper, we assume a rich set of fields, e.g. DNS response data. New architectures such as P4 [6] have programmable parsers that allow users to customize their applications to the set of fields required.
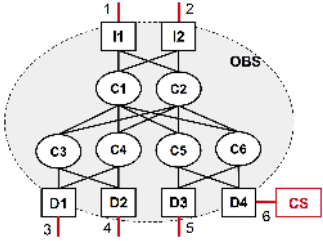

Figure 2: Topology for the running example.

network structure, and recompilation is needed only if the topology changes. By combining `DNS-tunnel-detect` with `assign-egress`, we have implemented a useful end-to-end program: `DNS-tunnel-detect;assign-egress`.

Monitoring. Suppose the operator wants to monitor packets entering the network at each ingress port (ports 1-6). She might use an array indexed by `inport` and increment the corresponding element on packet arrival: `count[inport]++`. Monitoring should take place *alongside* the rest of the program; thus, she might combine it using parallel composition (`+`): `(DNS-tunnel-detect + count[inport]++); assign-egress`. Intuitively, `p + q` makes a copy of the incoming packet and executes both `p` and `q` on it simultaneously.

Note that it is not always legal to compose two programs in parallel. For instance, if one writes to the same global variable that the other reads, there is a race condition, which leads to ambiguous state in the final program. Our compiler detects such race conditions and rejects ambiguous programs.

Network Transactions. Suppose that an operator sets up a honeypot at port 3 with IP subnet 10.0.3.0/25. The following program records, per inport, the IP and dstport of the last packet destined to the honeypot:

```
if dstip = 10.0.3.0/25
then hon-ip[inport] <- srcip;
     hon-dstport[inport] <- dstport
else id
```

Since this program processes many packets simultaneously, it has an implicit race condition: if packets p_1 and p_2, both destined to the honeypot, enter the network from port 1 and get reordered, each may visit `hon-ip` and `hon-dstport` in a different order (if the variables reside in different locations). Therefore, it is possible that `hon-ip[1]` contains the source IP of p_1 and `hon-dstport[1]` the destination port of p_2 while the operator's intention was that both variables refer to the same packet. To establish such properties for a collection of state variables, programmers can use *network transactions* by simply enclosing a series of statements in an `atomic` block. Atomic blocks co-locate their enclosed state variables so that a series of updates can be made to appear atomic.

2.2 Realizing Programs on the Data Plane

Consider `DNS-tunnel-detect; assign-egress`. To distribute this program across network devices, the SNAP

compiler should decide (i) where to place state variables (`orphan`, `susp-client`, and `blacklist`), and (ii) how packets should be routed across the physical network. These decisions should be made in such a way that each packet passes through devices storing *every* state variable it *needs*, in the correct *order*. Therefore, the compiler needs information about which packets need which state variables. In our example program, for instance, packets with `dstip = 10.0.6.0/24` and `srcport = 53` need to pass all three state variables, with `blacklist` accessed after the other two.

Program analysis. To extract the above information, we transform the program to an intermediate representation called *extended forwarding decision diagram (xFDD)* (see Figure 3). FDDs were originally introduced in an earlier work [35]. We extended FDDs in SNAP to support stateful packet processing. An xFDD is like a binary decision diagram: each intermediate node is a test on either packet fields or state variables. The leaf nodes are sets of action sequences, rather than merely 'true' and 'false' as in a BDD [1]. Each interior node has two successors: *true* (solid line), which determines the rest of the forwarding decision process for inputs passing the test, and *false* (dashed line) for failed cases. xFDDs are constructed compositionally; the xFDDs for different parts of the program are combined to construct the final xFDD. Composition is particularly more involved with stateful operations: the same state variable may be referenced in two xFDDs with different header fields, e.g., once as `s[srcip]` and then as `s[dstip]`. How can we know whether or not those fields are equal in the packet? We add a new kind of test, over pairs of packet fields (`srcip = dstip`), and new ordering requirements on the xFDD structure.

Once the program is transformed to an xFDD, we analyze the xFDD to extract information about which groups of packets need which state variables. In Figure 3, for example, leaf number 10 is on the true branch of `dstip=10.0.6.0/24` and `srcport=53`, which indicates that all packets with this property may end up there. These packets need `orphan`, because it is modified, and `susp-client`, because it is both tested and modified on the path. We can also deduce these packets can enter the network from any port and the ones that are not dropped will exit port 6. Thus, we can use the xFDD to figure out which packets need which state variables, aggregate this information across OBS ports, and choose paths for traffic between these ports accordingly.

Joint placement and routing. At this stage, the compiler has the information it needs to distribute the program. It uses a mixed-integer linear program (MILP) that solves an extension of the multi-commodity flow problem to *jointly* decide state placement and routing while minimizing network congestion. The constraints in the MILP guarantee that the selected paths for each pair of OBS ports take corresponding packets through devices storing every state variable that they need, in the correct order. Note that the xFDD
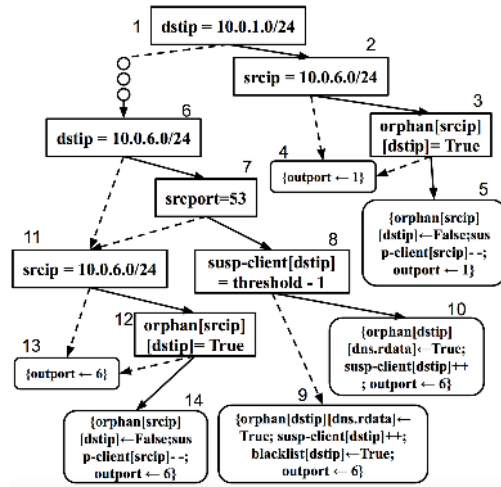

Figure 3: The equivalent xFDD for `DNS-tunnel-detect; assign-egress`

analysis can identify cases in which both directions of a connection need the same state variable s, so the MILP ensures they both traverse the device holding s.

In our example program, the MILP places all state variables on D_4, which is the optimal location as all packets to and from the protected subnet must flow through D_4.[2] Note that this is not obvious from the `DNS-tunnel-detect` code alone, but rather from its *combination* with `assign-egress`. This highlights the fact that in SNAP, program components can be written in a modular way, while the compiler makes globally optimal decisions using information from all parts. The optimizer also decides forwarding paths between external ports. For instance, traffic from I_1 and D_1 will go through C_1 and C_5 to reach D_4. The path from I_2 and D_2 to D_4 goes through C_2 and C_6, and D_3 uses C_5 to reach D_4. The paths between the rest of the ports are also determined by the MILP in a way that minimizes link utilization. The compiler takes state placement and routing results from the MILP, partitions the program's intermediate representation (xFDD) among switches, and generates rules for the controller to push to all stateless and stateful switches in the network.

Reacting to network events. The above phases only run if the operator changes the OBS program. Once the program compiles, and to respond to network events such as failures or traffic shifts, we use a simpler and much faster version of the MILP that given the current state placement, only re-optimizes for routing. Moreover, with state on the data plane, policy changes become considerably *less frequent* because the policy, and consequently switch configurations, *do not* change upon changes to state. In `DNS-tunnel-detect`, for instance, attack detection and mitigation are both captured in the program itself, happen *on the data plane*, and therefore react rapidly to malicious activities in the network.

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[2] State can be spread out across the network. It just happens that in this case, one location turns out to be optimal.

This is in contrast to the case where all the state is on the controller. There, the policy needs to change and recompile multiple times both during detection and on mitigation, to reflect the state changes on the controller in the rules on the data plane.

3. SNAP

SNAP is a high-level language with two key features: programs are *stateful* and are written in terms of an abstract network topology comprising a *one-big-switch* (OBS). It has an algebraic structure patterned on the NetCore/NetKAT family of languages [19, 2], with each program comprising one or more *predicates* and *policies*. SNAP's syntax is in Figure 4. Its semantics is defined through an evaluation function "eval." eval determines, in mathematical notation, how an input packet should be processed by a SNAP program. Note that this is part of the *specification* of the language, *not* the implementation. Any implementation of SNAP, including ours, should ensure that packets are processed as defined by the eval function: when we talk about "running" a program on a packet, we mean calling eval on that program and packet. We discuss eval's most interesting cases here; see the extended version for a full definition [42].

eval takes the SNAP term of interest, a packet, and a starting state and yields a set of packets and an output state. To properly define the semantics of multiple updates to state when programs are composed, we need to know the reads and writes to state variables performed by each program while evaluating the packet. Thus, eval also returns a *log* containing this information. It adds "$R\,s$" to the log whenever a read from state variable s occurs, and "$W\,s$" on writes. Note that these logs are part of our formalism, but not our implementation. We express the program state as a dictionary that maps state variables to their contents. The contents of each state variable is itself a mapping from values to values. Values are defined as packet-related fields (IP address, TCP ports, MAC addresses, DNS domains) along with integers, booleans and vectors of such values.

Predicates. Predicates have a constrained semantics: they never update the state (but may read from it), and either return the empty set or the singleton set containing the input packet. That is, they either pass or drop the input packet. *id* passes the packet and *drop* drops it. The test $f = v$ passes a packet *pkt* if the field f of *pkt* is v. These predicates yield empty logs.

The novel predicate in SNAP is the *state test*, written $s[e_1] = e_2$ and read "state variable (array) s at index e_1 equals e_2". Here e_1 and e_2 are *expressions*, where an expression is either a value v (like an IP address or TCP port), a field f, or a vector of them $\vec{e}$. For $s[e_1] = e_2$, function eval evaluates e_1 and e_2 on the input packet to yield two values v_1 and v_2. The packet can pass if state variable s indexed at v_1 is equal to v_2, and is dropped otherwise. The returned log will include $R\,s$, to record

Figure 4: SNAP's syntax. Highlighted items are not in NetCore.

that the predicate read from the state variable s.

We evaluate negation $\neg x$ by running eval on x and then complementing the result, propagating whatever log x produces. $x|y$ (disjunction) unions the results of running x and y individually, doing the reads of both x and y. $x\&y$ (conjunction) intersects the results of running x and y while doing the reads of x and then y.

Policies. Policies can modify packets and the state. Every predicate is a policy—it simply makes no modifications. Field modification $f \leftarrow v$ takes an input packet *pkt* and yields a new packet, *pkt′*, such that $pkt'.f = v$ but otherwise *pkt′* is the same as *pkt*. State update $s[e_1] \leftarrow e_2$ passes the input packet through while (i) updating the state so that s at eval(e_1) is set to eval(e_2), and (ii) adding $W\,s$ to the log. The $s[e]$++ (resp. --) operators increment (decrement) the value of $s[e]$ and add $W\,s$ to the log.

Parallel composition $p + q$ runs p and q in parallel and tries to merge the results. If the logs indicate a state read/write or write/write conflict for p and q then there is no consistent semantics we can provide, and we leave the semantics undefined. Take for example $(s[0] \leftarrow 1) + (s'[0] \leftarrow 2)$. There is no conflict if $s \neq s'$. However, the state updates conflict if $s = s'$. There is no good choice here, so we leave the semantics undefined and raise *compile error* in the implementation.

Sequential composition $p;q$ runs p and then runs q on each packet that p returned, merging the final results. We must ensure the runs of q are pairwise consistent, or else we will have a read/write or write/write conflict. For example, let p be $(f \leftarrow 1 + f \leftarrow 2)$, and $pkt[f \mapsto v]$ denote "update *pkt*'s f field to v". Given a packet *pkt*, the policy p produces two packets: $pkt_1 = pkt[f \mapsto 1]$ and $pkt_2 = pkt[f \mapsto 2]$. Let q be $s[0] \leftarrow f$, running $p;q$ fails because running q on pkt_1 and pkt_2 updates $s[0]$ differently. However, $p;q$ runs fine for $q = g \leftarrow 3$.

We have an explicit conditional "if a then p else q," which indicates *either* p or q are executed. Hence, both p and q can perform reads and writes to the same state.
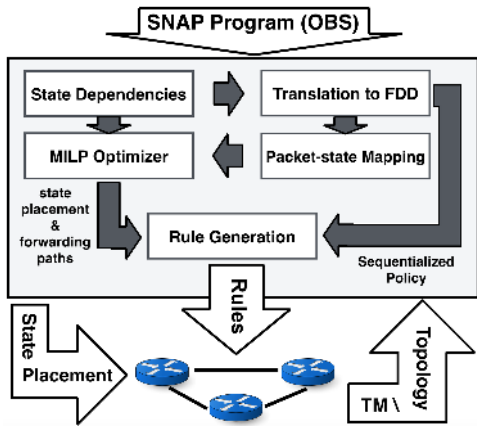
Figure 5: Overview of the compiler phases.

$$
\begin{array}{llll}
d & ::= & t \; ? \; d_1 : d_2 \mid \{as_1, \ldots, as_n\} & \text{xFDDs} \\
t & ::= & f = v \mid f_1 = f_2 \mid s[e_1] = e_2 & \text{tests} \\
as & ::= & a \mid a; a & \text{action sequences} \\
a & ::= & id \mid drop \mid f \leftarrow v \mid s[e_1] \leftarrow e_2 & \text{actions} \\
& & \mid s[e_1]\texttt{++} \mid s[e_1]\texttt{--}
\end{array}
$$

$$
\begin{aligned}
\text{TO-XFDD}(a) &= \{a\} \\
\text{TO-XFDD}(f = v) &= f = v \; ? \; \{id\} \; : \; \{drop\} \\
\text{TO-XFDD}(\neg x) &= \ominus \text{TO-XFDD}(x) \\
\text{TO-XFDD}(s[e_1] = e_2) &= s[e_1] = e_2 \; ? \; \{id\} \; : \; \{drop\} \\
\text{TO-XFDD}(\mathsf{atomic}(p)) &= \text{TO-XFDD}(p) \\
\text{TO-XFDD}(p + q) &= \text{TO-XFDD}(p) \oplus \text{TO-XFDD}(q) \\
\text{TO-XFDD}(p; q) &= \text{TO-XFDD}(p) \odot \text{TO-XFDD}(q) \\
\text{TO-XFDD}(\text{if } x \text{ then } p \text{ else } q) &= (\text{TO-XFDD}(x) \odot \text{TO-XFDD}(p)) \\
&\quad \oplus \; (\ominus \text{TO-XFDD}(x) \odot \text{TO-XFDD}(q))
\end{aligned}
$$

Figure 6: xFDD syntax and translation.

We have a notation for *atomic blocks*, written $\mathsf{atomic}(p)$. As described in §2, there is a risk of inconsistency between state variables residing on different switches on a real network when many packets are in flight concurrently. When compiling $\mathsf{atomic}(p)$, our compiler ensures that all the state in p is updated atomically (§4).

4. COMPILATION

To implement a SNAP program specified on one big switch, we must fill in two critical details: *traffic routing* and *state placement*. The physical topology may offer many paths between edge ports, and many possible locations for placing state.[3] The routing and placement problems interact: if two flows (with different input and output OBS ports) both need some state variable s, we should select routes for the two flows such that they pass through a common location where we place s. Further complicating the situation, the OBS program may specify that certain flows read/write multiple state variables in a particular order. The routing and placement on the physical topology must respect that order. In DNS-tunnel-detect, for instance, routing must ensure that packets reach wherever orphan is placed before susp-client. In some cases, two different flows may depend on the same state variables, but in different orders.

We have designed a *compiler* that translates OBS programs into forwarding rules and state placements for a given topology. As shown in Figure 5, the two key phases are (i) translation to *extended forwarding decision diagrams* (xFDDs)—used as the intermediate representation of the program and to calculate which flows need which state variables—and (ii) optimization via *mixed integer linear program* (MILP)—used to decide routing and state placement. In the rest of this section, we present the compilation process in phases, first discussing the analysis of state dependencies, followed by the translation to xFDDs and the packet-state mapping, then the optimization problems, and finally

[3] In this work, we assume each state variable resides in one place, though it is conceivable to distribute it (see §4.4 and §7.3).

the generation of rules sent to the switches.

4.1 State Dependency Analysis

Given a program, the compiler first performs *state dependency analysis* to determine the ordering constraints on its state variables. A state variable t *depends* on a state variable s if the program writes to t after reading from s. Any realization of the program on a concrete network must ensure that t does not come before s. Parallel composition, $p + q$, introduces no dependencies: if p reads or writes state, then q can run independently of that. Sequential composition $p; q$, on the other hand, introduces dependencies: whatever reads are in p must happen before writes in q. In explicit conditionals "if a then p else q", the writes in p and q depend on the condition a. Finally, atomic sections $\mathsf{atomic}(p)$ say that all state in p is inter-dependent. In DNS-tunnel-detect, for instance, blacklist is dependent on susp-client, itself dependent on orphan. This information is encoded as a dependency graph on state variables and is used to order the xFDD structure (§4.2), and in the MILP (§4.4) to drive state placement.

4.2 Extended Forwarding Decision Diagrams

The input to the compiler is a SNAP program, which can be a composition of several smaller programs. The output, on the other end, is the distribution of the original policy across the network. Thus, in between, we need an intermediate representation for SNAP programs that is both composable and easily partitioned. This intermediate representation can help the compiler compose small program pieces into a unified representation, which can further be partitioned to get distributed across the network. Extended forwarding decision diagrams (xFDDs), which are introduced in this section, are what we use as our internal representation of SNAP programs and have both desired properties. They also simplify analysis of SNAP programs for extracting packet-state mapping, which we discuss in §4.3

Formally (see Figure 6), an xFDD is either a *branch* $(t \; ? \; d_1 : d_2)$, where t is a test and d_1 and d_2 are xFDDs, or a *set* of action sequences $\{as_1, \ldots, as_n\}$. Each branch

$$\{as_{11}, \cdots, as_{1n}\} \oplus \{as_{21}, \cdots, as_{2m}\} = \{as_{11}, \cdots, as_{1n}\} \cup \{as_{21}, \cdots, as_{2m}\}$$
$$(t \; ? \; d_1 : d_2) \oplus \{as_1, \cdots, as_n\} = (t \; ? \; d_1 \oplus \{as_1, \cdots, as_n\} : d_2 \oplus \{as_1, \cdots, as_n\})$$

$$(t_1 \; ? \; d_{11} : d_{12}) \oplus (t_2 \; ? \; d_{21} : d_{22}) = \begin{cases} (t_1 \; ? \; d_{11} \oplus d_{21} : d_{12} \oplus d_{22}) & t_1 = t_2 \\ (t_1 \; ? \; d_{11} \oplus (t_2 \; ? \; d_{21} : d_{22}) : d_{12} \oplus (t_2 \; ? \; d_{21} : d_{22}) & t_1 \sqsubset t_2 \\ (t_2 \; ? \; d_{21} \oplus (t_1 \; ? \; d_{11} : d_{12}) : d_{22} \oplus (t_1 \; ? \; d_{11} : d_{12}) & t_2 \sqsubset t_1 \end{cases}$$

$$\ominus \{id\} = \{drop\}$$
$$\ominus \{drop\} = \{id\}$$
$$\ominus (t?d_1 : d_2) = (t? \ominus d_1 : \ominus d_2)$$

$$\begin{aligned} as \odot \{as_1, \cdots, as_n\} &= \{as \odot as_1, \cdots, as \odot as_n\} \\ as \odot (t \; ? \; d_1 : d_2) &= (\text{see explanations in §4.2}) \\ \{as_1, \cdots, as_n\} \odot d &= (as_1 \odot d) \oplus \cdots \oplus (as_n \odot d) \\ (t \; ? \; d_1 : d_2) \odot d &= (d_1 \odot d)|_t \oplus (d_2 \odot d)|_{\sim t} \end{aligned}$$

$$\{as_1, \cdots, as_n\}|_t = (t \; ? \; \{as_1, \cdots, as_n\} : \{drop\})$$

$$(t_1 \; ? \; d_1 : d_2)|_{t_2} = \begin{cases} (t_1 \; ? \; d_1 : \{drop\}) & t_1 = t_2 \\ (t_2 \; ? \; (t_1 \; ? \; d_1 : d_2) : \{drop\}) & t_2 \sqsubset t_1 \\ (t_1 \; ? \; d_1|_{t_2} : d_2|_{t_2}) & t_1 \sqsubset t_2 \end{cases}$$

Figure 7: Definitions of xFDD composition operators.

can be thought of as a conditional: if the test t holds on a given packet pkt, then the xFDD continues processing pkt using d_1; if not, processes pkt using d_2. There are three kinds of tests. The *field-value test* $f = v$ holds when $pkt.f$ is equal to v. The *field-field test* $f_1 = f_2$ holds when the values in $pkt.f_1$ and $pkt.f_2$ are equal. Finally, the *state test* $s[e_1] = e_2$ holds when the state variable s at index e_1 is equal to e_2. The last two tests are our extensions to FDDs. The state tests support our stateful primitives, and as we show later in this section, the field-field tests are required for correct compilation. Each leaf in an xFDD is a set of action sequences, with each action being either the identity, drop, field-update $f \leftarrow v$, or state update $s[e_1] \leftarrow e_2$, which is another extension to the original FDD.

A key property of xFDDs is that the order of their tests ($\sqsubset$) must be defined in advance. This ordering is necessary to ensure that each test is present at most once on any path in the final tree when merging two xFDDs into one. Thus, xFDD composition can be done efficiently without creating redundant tests. In our xFDDs, we ensure that all field-value tests precede all field-field tests, themselves preceding all state tests. Field-value tests themselves are ordered by fixing an arbitrary order on fields and values. Field-field tests are ordered similarly. For state tests, we first define a total order on state variables by looking at the dependency graph from §4.1. We break the dependency graph into strongly connected components (SCCs) and fix an arbitrary order on state variables within each SCC. For every edge from one SCC to another, i.e., where some state variable in the second SCC depends on some state variable in the first, s_1 precedes s_2 in the order, where s_2 is the minimal element in the second SCC and s_1 is the maximal element in the first SCC. The state tests are then ordered based on the order of state variables.

We translate a program to an xFDD using the TO-XFDD function (Figure 6), which translates small parts of a program directly to xFDDs. Composite programs get recursively translated and then composed using a corresponding composition operator for xFDDs: we use $\oplus$ for $p + q$, $\odot$ for $p \; ; \; q$, and $\ominus$ for $\neg p$. Figure 7 gives a high-level definition of the semantics of these operators. For example, $d_1 \oplus d_2$ tries to merge similar test nodes recursively by merging their true branches together and false ones together. If the two tests are

not the same and d_1's test comes first in the total order, both of its subtrees are merged recursively with d_2. The other case is similar. $d_1 \oplus d_2$ for leaf nodes is the union of their action sets.

The hardest case is surely for $\odot$, where we try to add in an action sequence as to an xFDD $(t \; ? \; d_1 : d_2)$. Suppose we want to compose $f \leftarrow v_1$ with $(f = v_2 \; ? \; d_1 : d_2)$. The result of this xFDD composition should behave as if we first do the update and then the condition on f. If $v_1 = v_2$, the composition should continue only on d_1, and if not, only on d_2. Now let's look at a similar example including state, composing $s[srcip] \leftarrow e_1$ with $(s[dstip] = e_2 \; ? \; d_1 : d_2)$. If $srcip$ and $dstip$ are equal (rare but not impossible) and e_1 and e_2 always evaluate to the same value, then the whole composition reduces to just d_1. The field-field tests are introduced to let us answer these equality questions, and that is why they always precede state tests in the tree. The trickiness in the algorithm comes from generating proper field-field tests, by keeping track of the information in the xFDD, to properly answer the equality tests of interest. The full algorithm is given in the extended version [42].

Inconsistent use of state variables is prohibited by the language semantics when composing programs (see §3). We enforce the semantics by looking for these violations while merging the xFDDs of composed programs and raising a compile error if the final xFDD contains a leaf with parallel updates to the same state variable.

4.3 Packet-State Mapping

For a given program p, the corresponding xFDD d offers an explicit and complete specification of the way p handles packets. We analyze d, using an algorithm called *packet-state mapping*, to determine which *flows* use which states. This information is further used in the optimization problem (§4.4) to decide the correct routing for each flow. Our default definition of a flow is those packets that travel between any given pair of ingress/egress ports in the OBS, though we can use other notions of flow (see §4.4). Traversing from d's root down to the action sets at d's leaves, we can gather information associating each flow with the set of state variables read or written. We omit the full algorithm due to space constraints.

Furthermore, the operators can give hints to the compiler by specifying their network *assumptions* in a sep-

arate policy:

```
assumption = (srcip = 10.0.1.0/24 & inport = 1)
           + (srcip = 10.0.2.0/24 & inport = 2)
           + ...
           + (srcip = 10.0.6.0/24 & inport = 6)
```

We require the assumption policy to be a predicate over packet header fields, only passing the packets that match the operator's assumptions. `assumption` is then sequentially composed with the rest of the program, enforcing the assumption by dropping packets that do not match the assumption. Such assumptions benefit the packet-state mapping. Consider our example xFDD in Figure 3. Following the xFDD's tree structure, we can infer that all the packets going to port 6 need all the three state variables in `DNS-tunnel-detect`. We can also infer that all the packets coming from the 10.0.6.0/24 subnet need `orphan` and `susp-client`. However, there is nothing in the program to tell the compiler that these packets can only enter the network from port 6. Thus, the above assumption policy can help the compiler to identify this relation and place state more efficiently.

4.4 State Placement and Routing

At this stage, the compiler has enough information to fill in the details abstracted away from the programmer: where and how each state variable should be placed, and how the traffic should be routed in the network. There are two general approaches for deciding state placement and routing. One is to keep *each* state variable at one location and route the traffic through the state variables it needs. The other is to keep multiple copies of the same state variable on different switches and partition and route the traffic through them. The second approach requires mechanisms to keep different copies of the same state variable consistent. However, it is not possible to provide strong consistency guarantees when distributed updates are made on a packet-by-packet basis at line rate. Therefore, we chose the first approach, which locates each state variable at one physical switch.

To decide state placement and routing, we generate an optimization problem, a *mixed-integer linear program* (MILP) that is an extension of the multi-commodity flow linear program. The MILP has three key inputs: the concrete network topology, the state dependency graph G, and the packet-state mapping, and two key outputs: routing and state placement (Table 1). Since route selection depends on state placement and each state variable is constrained to one physical location, we need to make sure the MILP picks *correct* paths without degrading network performance. Thus, the MILP minimizes the sum of link utilization in the network as a measure of congestion. However, other objectives or constraints are conceivable to customize the MILP to other kinds of performance requirements.
Inputs. The topology is defined in terms of the following inputs to the MILP: (i) the nodes, some distinguished as edges (ports in OBS), (ii) expected traffic

Variable	Description
u, v	edge nodes (ports in OBS)
n	physical switches in the network
i, j	all nodes in the network
d_{uv}	traffic demand between u and v
c_{ij}	link capacity between i and j
dep	state dependencies
$tied$	co-location dependencies
S_{uv}	state variables needed for flow uv
R_{uvij}	fraction of d_{uv} on link (i, j)
P_{sn}	1 if state s is placed on n, 0 otherwise
P_{suvij}	d_{uv} fraction on link (i, j) that has passed s

Table 1: Inputs and outputs of the optimization problem.

d_{uv} for every pair of edge nodes u and v, and (iii) link capacities c_{ij} for every pair of nodes i and j. State dependencies in G are translated into input sets dep and *tied*. *tied* contains pairs of state variables which are in the same SCC in G, and must be co-located. *dep* identifies state variables with dependencies that do not need to be co-located; in particular, $(s, t) \in dep$ when s precedes t in variable ordering, and they are not in the same SCC in G. The packet-state mapping is used as the input variables S_{uv}, identifying the set of state variables needed on flows between nodes u and v.
Outputs and Constraints. The routing outputs are variables R_{uvij}, indicating what fraction of the flow from edge node u to v should traverse the link between nodes i and j. The constraints on R_{uvij} (left side of Table 2) follow the multi-commodity flow problem closely, with standard link capacity and flow conservation constraints, and edge nodes distinguished as sources and sinks of traffic.

State placement is determined by the variables P_{sn}, which indicate whether the state variable s should be placed on the physical switch n. Our constraints here are more unique to our setting. First, every state variable s can be placed on exactly one switch, a choice we discussed earlier in this section. Second, we must ensure that flows that need a given state variable s traverse that switch. Third, we must ensure that each flow traverses states in the order specified by the *dep* relation; this is what the variables P_{suvij} are for. We require that $P_{suvij} = R_{uvij}$ when the traffic from u to v that goes over the link (i, j) has already passed the switch with the state variable s, and zero otherwise. If *dep* requires that s should come before some other state variable t— and if the (u, v) flow needs both s and t—we can use P_{suvij} to make sure that the (u, v) flow traverses the switch with t only after it has traversed the switch with s (the last state constraint in Table 2). Finally, we must make sure that state variables $(s, t) \in tied$ are located on the same switch. Note that only state variables that are *inter-dependent* are required to be located on the same switch. Two variables s and t are inter-dependent if a read from s is required before a write to t *and vice versa*. Placing them on different switches will result in a forwarding loop between the two switches which is not desirable in most networks. Therefore, in order to syn-

Routing Constraints	State Constraints
	$\sum_n P_{sn} = 1$
$\sum_j R_{uvuj} = 1$	$\forall u, v. \ \forall s \in S_{uv}. \ \sum_i R_{uvin} \geq P_{sn}$
$\sum_i R_{uviv} = 1$	$\forall (s,t) \in tied. \ P_{sn} = P_{tn}$
$\sum_{u,v} R_{uvij} d_{uv} \leq c_{ij}$	$P_{suvij} \leq R_{uvij}$
$\sum_i R_{uvin} = \sum_j R_{uvnj}$	$P_{sn} + \Sigma_i P_{suvin} = \Sigma_j P_{suvnj}$
$\sum_i R_{uvin} \leq 1$	$\forall s \in S_{uv}. \ P_{sv} + \sum_i P_{suviv} = 1$
	$P_{sn} + \Sigma_i P_{suvin} \geq P_{tn}$

Table 2: Constraints of the optimization problem.

chronize reads and writes to inter-dependent variables correctly, they are always placed on the same switch.

Although the current prototype chooses the same path for the traffic between the same ports, the MILP can be configured to decide paths for more fine-grained notions of flows. Suppose packet-state mapping finds that only packets with $srcip = x$ need state variable s. We refine the MILP input to have two edge nodes per port, one for traffic with $srcip = x$ and one for the rest, so the MILP can choose different paths for them.

Finally, the MILP makes a *joint* decision for state placement and routing. Therefore, path selection is tied to state placement. To have more freedom in picking forwarding paths, one option is to first use common traffic engineering techniques to decide routing, and then optimize the placement of state variables with respect to the selected paths. However, this approach may require replicating state variables and maintaining consistency across multiple copies, which as mentioned earlier, is not possible at line rate for distributed packet-by-packet updates to state variables.

4.5 Generating Data-Plane Rules

Rule generation happens in two phases and combines information from the xFDD and MILP to configure the network switches. We assume each packet is augmented with a SNAP-header upon entering the network, which contains its original OBS inport and future outport, and the id of the last processed xFDD node, the purpose of which will be explained shortly. This header is stripped off by the egress switch when the packet exits the network. We use `DNS-tunnel-detect;assign-egress` from §2 as a running example, with its xFDD in Figure 3. For the sake of the example, we assume that all the state variables are stored on C_6 instead of D_4.

In the first phase, we break the xFDD down into 'per-switch' xFDDs, since not every switch needs the entire xFDD to process packets. Splitting the xFDD is straightforward given placement information: stateless tests and actions can happen anywhere, but reads and writes of state variables must happen on switches storing them. For example, edge switches (I_1 and I_2, and D_1 to D_4) only need to process packets up to the state tests, e.g., tests 3 and 8, and write the test number in the packet's SNAP-header showing how far into the xFDD they progressed. Then, they send the packets to C_6, which has the corresponding state variables, `orphan` and `susp-client`. C_6, on the other hand, does not need

the top part of the xFDD. It just needs the subtrees containing its state variables to continue processing the packets sent from the edges. The per-switch xFDDs are then translated to switch-level configurations, by a straightforward traversal of the xFDD (See §5).

In the second phase, we generate a set of match-action rules that take packets through the paths decided by the MILP. These paths comply with the state ordering used in the xFDD, thus they get packets to switches with the right states in the right order. Note that packets contain the path identifier (the OBS inport and outport, (u, v) pair in this case) and the "routing" match-action rules are generated in terms of this identifier to forward them on the correct path. Additionally, note that it may not always be possible to decide the egress port v for a packet upon entry if its outport depends on state. We observe that in that case, all the paths for possible outports of the packet pass the state variables it needs. We load-balance over these paths in proportion to their capacity and show, in the extended version [42], that traffic on these paths remains in their capacity limit.

To see an example of how packets are handled by generated rules, consider a DNS response with source IP 10.0.1.1 and destination IP 10.0.6.6, entering the network from port 1. The rules on I_1 process the packet up to test 8 in the xFDD, tag the packet with the path identifier (1, 6) and number 8. The packet is then sent to C_6. There, C_6 will process the packet from test 8, update state variables accordingly, and send the packet to D_4 to exit the network from port 6.

5. IMPLEMENTATION

The compiler is mostly implemented in Python, except for the state placement and routing phase (§4.4) which uses the Gurobi Optimizer [15] to solve the MILP. The compiler's output for each switch is a set of switch-level instructions in a low-level language called NetASM [32], which comes with a software switch capable of executing those instructions. NetASM is an assembly language for programmable data planes designed to serve as the "narrow waist" between high-level languages such as SNAP, and NetCore[19], and programmable switching architectures such as RMT [7], FPGAs, network processors and Open vSwitch.

As described in §4.5, each switch processes the packet by its customized per-switch xFDD, and then forwards it based on the fields of the SNAP-header using a match-action table. To translate the switch's xFDD to NetASM instructions, we traverse the xFDD and generate a *branch* instruction for each test node, which jumps to the instruction of either the true or false branch based on the test's result. Moreover, we generate instructions to create two tables for each state variable, one for the indices and one for the values. In the case of a state test in the xFDD, we first retrieve the value corresponding to the index that matches the packet, and then perform the branch. For xFDD leaf nodes, we generate *store*

	Application
Chimera [5]	# domains sharing the same IP address
	# distinct IP addresses under the same domain
	DNS TTL change tracking
	DNS tunnel detection
	Sidejack detection
	Phishing/spam detection
FAST [21]	Stateful firewall
	FTP monitoring
	Heavy-hitter detection
	Super-spreader detection
	Sampling based on flow size
	Selective packet dropping (MPEG frames)
	Connection affinity
Bohatei [8]	SYN flood detection
	DNS amplification mitigation
	UDP flood mitigation
	Elephant flows detection
Others	Bump-on-the-wire TCP state machine
	Snort flowbits [36]

Table 3: Applications written in SNAP.

ID	Phase		Topo/TM Change	Policy Change	Cold Start
P1	State dependency		-	✓	✓
P2	xFDD generation		-	✓	✓
P3	Packet-state map		-	✓	✓
P4	MILP creation		-	-	✓
P5	MILP solving	State placement and routing (ST)	-	✓	✓
		Routing (TE)	✓	-	-
P6	Rule generation		✓	✓	✓

Table 4: Compiler phases. For each scenario, phases that get executed are checkmarked.

Topology	# Switches	# Edges	# Demands
Stanford	26	92	20736
Berkeley	25	96	34225
Purdue	98	232	24336
AS 1755	87	322	3600
AS 1221	104	302	5184
AS 6461	138	744	9216
AS 3257	161	656	12544

Table 5: Statistics of evaluated enterprise/ISP topologies.

instructions that modify the packet fields and state tables accordingly. Finally, we use NetASM support for atomic execution of multiple instructions to guarantee that operations on state tables happen atomically.

While NetASM was useful for testing our compiler, any programmable device that supports match-action tables, branch instructions, and stateful operations can be a SNAP target. The prioritized rules in match-action tables, for instance, are effectively branch instructions. Thus, one can use multiple match-action tables to implement xFDD in the data plane, generating a separate rule for each path in the xFDD. Several emerging switch interfaces support stateful operations [6, 4, 38, 25]. We discuss possible software and hardware implementations for SNAP stateful operations in §7.

6. EVALUATION

This section evaluates SNAP in terms of language expressiveness and compiler performance.

6.1 Language Expressiveness

We have implemented several stateful network functions (Table 3) that are typically relegated to middleboxes in SNAP. Examples were taken from the Chimera [5], FAST [21], and Bohatei [8] systems. The code can be found in the extended version [42]. Most examples use protocol-related fields in fixed packet-offset locations, which are parsable by emerging programmable parsers. Some fields require session reassembly. However, this is orthogonal to the language expressiveness; as long as these fields are available to the switch, they can be used in SNAP programs. To make them available, one could extract these fields by placing a "preprocessor" before the switch pipeline, similar to middleboxes. For instance, Snort [36] uses preprocessors to extract fields for use in the detection engine.

6.2 Compiler Performance

The compiler goes through several phases upon the system's cold start, yet most events require only some

of them. Table 4 summarizes these phases and their sensitivity to network and policy changes.

Cold Start. When the very first program is compiled, the compiler goes through all phases, including MILP model creation, which happens *only once* in the lifetime of the network. Once created, the model supports incremental additions and modifications of variables and constraints in a few milliseconds.

Policy Changes. Compiling a *new* program requires executing the three program analysis phases and rule generation as well as *both* state placement and routing, which are decided using the MILP in §4.4, denoted by "ST". Policy changes become considerably *less frequent* (§2.2) since most dynamic changes are captured by the state variables that reside on the data plane. The policy, and consequently switch configurations, *do not* change upon state changes. Thus, we expect policy changes to happen infrequently, and be planned in advance. The Snort rule set, for instance, gets updated every few days [37].

Topology/TM Changes. Once the policy is compiled, we fix the decided state placement, and only re-optimize routing in response to network events such as failures. For that, we formulated a variant of ST, denoted as "TE" (traffic engineering), that receives state placement as input, and decides forwarding paths while satisfying state requirement constraints. *We expect TE to run every few minutes* since in a typical network, the traffic matrix is fairly stable and traffic engineering happens on the timescale of *minutes* [16, 43, 22, 41].

6.2.1 Experiments

We evaluated performance based on applications listed in Table 3. Traffic matrices are synthesized using a gravity model [31]. We used an Intel Xeon E3, 3.4 GHz, 32GB server, and PyPy compiler [27].

Topologies. We used a set of three campus networks and four inferred ISP topologies from RocketFuel [40]
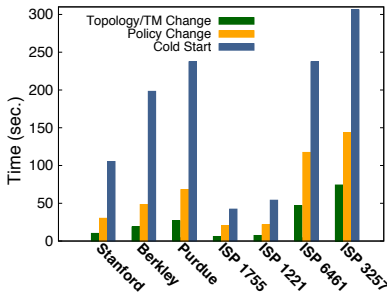
Figure 8: Compilation time of `DNS-tunnel-detect` with routing on enterprise/ISP networks.
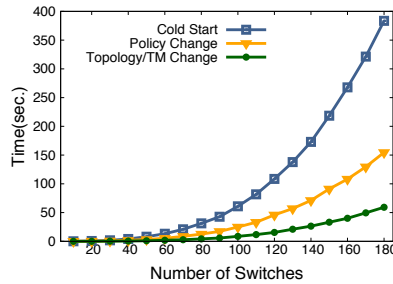
Figure 9: Compilation time of `DNS-tunnel-detect` with routing on IGen topologies.
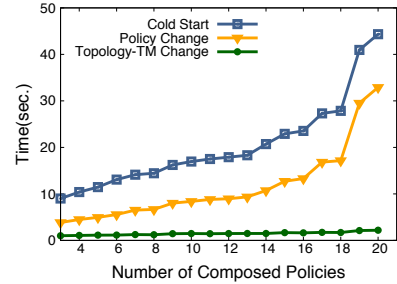
Figure 10: Compilation time for policies from Table 3 incrementally composed on a 50-switch network.

Figure 11: Compiler runtimes for scenarios in Table 4 on various policies and topologies. Once compiled for the first time (cold start, policy change), a policy reacts to traffic using its state variables. Topology/TM changes result in reoptimizing forwarding paths.

	P1-P2-P3 (s)	**P5** (s)		**P6**(s)	**P4** (s)
		ST	**TE**		
Stanford	1.1	29	10	0.1	75
Berkeley	1.5	47	18	0.1	150
Purdue	1.2	67	27	0.1	169
AS 1755	0.6	19	6	0.04	22
AS 1221	0.7	21	7	0.04	32
AS 6461	0.8	116	47	0.1	120
AS 3257	0.9	142	74	0.2	163

Table 6: Runtime of compiler phases when compiling `DNS-tunnel-detect` with routing on enterprise/ISP topologies.

(Table 5).[4] For ISP networks, we considered 70% of the switches with the lowest degrees as edge switches to form OBS external ports. The "# Demands" column shows the number of distinct OBS ingress/egress pairs. We assume directed links. Table 6 shows compilation time for the DNS tunneling example (§2) on each network, broken down by compiler phase. Figure 8 compares the compiler runtime for different scenarios, combining the runtimes of phases relevant for each.

Scaling with topology size. We synthesize networks with 10–180 switches using IGen [29]. In each network, 70% of the switches with the lowest degrees are chosen as edges and the DNS tunnel policy is compiled with that network as a target. Figure 9 shows the compilation time for different scenarios, combining the runtimes of phases relevant for each. Note that by increasing the topology size, the policy size also increases in the `assign-egress` and `assumption` parts.

Scaling with number of policies. The performance of several phases of the compiler, specially xFDD generation, is a function of the size and complexity of the input policy. Therefore, we evaluated how the compiler's performance scales with policy size using the example programs from Table 3. Given that these programs are taken from recent papers and tools in the literature [5, 21, 8, 36], we believe they form a fair benchmark for our

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[4] The publicly available Mininet instance of Stanford campus topology has 10 extra dummy switches to implement multiple links between two routers.

evaluation. Except for TCP state machine, the example programs are similar in size and complexity to the DNS tunnel example (§2). We use the 50-switch network from the previous experiment and start with the first program in Table 3. We then gradually increase the size of the final policy by combining this program with more programs from Table 3 using the parallel composition operator. Each additional component program affects traffic destined to a separate egress port.

Figure 10 depicts the compilation time as a function of the number of components from Table 3 that form the final policy. The 10-second jump from 18 to 19 takes place when the TCP state machine policy is added, which is considerably more complex than others. The increase in the compilation time mostly comes from the xFDD generation phase. In this phase, the composed programs are transformed into separate xFDDs, which are then combined to form the xFDD for the whole policy (§4.2). The cost of xFDD composition depends on the size of the operands, so as more components are put together, the cost grows. The cost may also depend on the order of xFDD composition. Our current prototype composes xFDDs in the same order as the programs themselves are composed and leaves finding the optimal order to compose xFDDs to future work.

The last data point in Figure 10 shows the compilation time of a policy composed of all the 20 examples in Table 3, with a total of 35 state variables. These policies are composed using parallel composition, which does not introduce read/write dependencies between state variables. Thus, the dependency graph for the final policy is a collection of the dependency graphs of the composed policies. Each of the composed policies affects the traffic to a separate egress port, which is detected by the compiler in the packet-state mapping phase. Thus, when compiled to the 50-switch network, state variables for each policy are placed on the switch closest to the egress port whose traffic the policy affects. If a policy were to affect a larger portion of traffic, e.g., the traffic of a set of ingress/egress ports, SNAP would place state variables in an optimal location where the

aggregated traffic of interest is passing through.

6.2.2 Analysis of Experimental Results

Creating the MILP takes longer than solving it, in most cases, and much longer than other phases. Fortunately, this is a *one-time* cost. After creating the MILP instance, incrementally adding or removing variables and constraints (as the topology and/or state requirements change) takes just a few milliseconds.

Solving the ST MILP unsurprisingly takes longer as compared to the rest of the phases when topology grows. It takes ∼ 2.5 minutes for the biggest synthesized topology and ∼ 2.3 minutes for the biggest RocketFuel topology. The curve is close to exponential as the problem is inherently computationally hard. However, this phase takes place only in cold start or upon a *policy* change, which are infrequent and planned in advance.

Re-optimizing routing with fixed state placement is much faster. In response to network events (e.g., link failures), TE MILP can recompute paths in around a minute across all our experiments, *which is the timescale we initially expected for this phase* as it runs in the topology/TM change scenarios. Moreover, it can be used even on *policy* changes, if the user settles for a sub-optimal state placement using heuristics rather than ST MILP. We plan to explore such heuristics.

Given the kinds of events that require complete (policy change) or partial (network events) recompilation, we believe that our compilation techniques meet the requirements of enterprise networks and medium-size ISPs. Moreover, if needed, our compilation procedure could be combined with traffic-engineering techniques once the state placement is decided, to avoid re-solving the original or even TE MILP on small timescales.

7. DISCUSSION

This section discusses data-plane implementation strategies for SNAP's stateful operations, how SNAP relates to middleboxes, and possible extensions to our techniques to enable a broader range of applications.

7.1 Stateful Operations in the Data Plane

A state variable (array) in SNAP is a key-value mapping, or a *dictionary*, on header fields, persistent across multiple packets. When the key (index) range is small, it is feasible to pre-allocate all the memory the dictionary needs and implement it using an array. A large but *sparse* dictionary can be implemented using a *reactively*-populated table, similar to a MAC learner table. It contains a single default entry in the beginning, and as packets fly by and change the state variable, it *reactively* adds/updates the corresponding entries.

In software, there are efficient techniques to implement a dictionary in either approach, and some software switches already support similar reactive "learning" operations, either atomically [32] or with small periods of inconsistency [25]. The options for current hardware

are: (i) arrays of registers, which are already supported in emerging switch interfaces [6]. They can be used to implement small dictionaries, as well as Bloom Filters and hash tables as sparse dictionaries. In the latter case, it is possible for two different keys to hash to the same dictionary entry. However, there are applications such as load balancing and flow-size-based sampling that can tolerate such collisions [21]. (ii) Content Addressable Memories (CAMs) are typically present in today's hardware switches and can be modified by a software agent running on the switch. Since CAM updates triggered by a packet are not immediately available to the following packets, it may be used for applications that tolerate small periods of state inconsistency, such as a MAC learner, DNS tunnel detection, and others from Table 3. Our NetASM implementation (§5) takes the CAM-based approach. NetASM's software switch supports atomic updates to the tables in the data plane and therefore can perform *consistent* stateful operations.

At the time of writing this paper, we are not aware of any hardware switch that can implement an *arbitrary* number of SNAP's stateful operations both at *line rate* and with *strong consistency.* Therefore, we use NetASM's low-level primitives as the compiler's back-end so that we can specify data-plane primitives that are required for an efficient and consistent implementation of SNAP's operations. If one is willing to relax one of the above constraints for a specific application, i.e., operating at line rate or strong consistency, it would be possible to implement SNAP on today's switches. If strong consistency is relaxed, CAMs/TCAMs can be programmed using languages such as P4 [6] to implement SNAP's stateful operations as described above. If line-rate processing is relaxed, one can use software switches, or programmable hardware switching devices such as ones in the OpenNFP project that allow insertion of Micro-C code extensions to P4 programs at the expense of processing speed [23] or FPGAs.

7.2 SNAP and Middleboxes

Networks traditionally rely on middleboxes for advanced packet processing, including stateful functionalities. However, advances in switch technology enable stateful packet processing in the data plane, which naturally makes the switches capable of subsuming a subset of middlebox functionality. SNAP provides a *high-level programming framework* to exploit this ability, hence, it is able to express a wide range of stateful programs that are typically relegated to middleboxes (see Table 3 for examples). This helps the programmer to think about a single, explicit network policy, as opposed to a disaggregated, implicit network policy using middleboxes, and therefore, get more control and customization over a variety of simpler stateful functionalities.

This also makes SNAP subject to similar challenges as managing stateful middleboxes. For example, many network functions must observe all traffic pertaining to a connection *in both directions*. In SNAP, if traf-

fic in both directions uses a shared state variable, the MILP optimizer forces traffic in both directions through the same node. Moreover, previous work such as Split/Merge [30] and OpenNF [13] show how to migrate *internal* state from one network function to another, and Gember-Jacobson et al. [12] manage to migrate state without buffering packets at the controller. SNAP currently focuses on static state placement. However, since SNAP's state variables are explicitly declared as part of the policy, rather than hidden inside blackbox software, SNAP is well situated to adopt these algorithms to support smooth transitions of state variables in dynamic state placement. Additionally, the SNAP compiler can easily analyze a program to determine whether a switch modifies packet fields to ensure correct traffic steering—something that is challenging today with blackbox middleboxes [9, 28].

While SNAP goes a step beyond previous high-level languages to incorporate stateful programming into SDN, we neither claim that it is as expressive as all stateful middleboxes, nor that it can replace them. To interact with middleboxes, SNAP may adopt techniques such as FlowTags [9] or SIMPLE [28] to direct traffic through middleboxes chains by tagging packets to mark their progress. Since SNAP has its own tagging and steering to keep track of the progress of packets through the policy's xFDD, this adoption may require integrating tags in the middlebox framework with SNAP's tags. As an example, we will describe below how SNAP and FlowTags can be used together on the same network.

In FlowTags, users specify which class of traffic should pass which chain of middleboxes under what conditions. For instance, they can ask for web traffic to go to an intrusion detection system (IDS) after a firewall if the firewall marks the traffic as suspicious. The controller keeps a mapping between the tags and the flow's original five tuple plus the contextual information of the last middlebox, e.g., suspicious vs. benign in the case of a firewall. The tags are used for steering the traffic through the right chain of middleboxes and preserving the original information of the flow in case it is changed by middleboxes. To use FlowTags with SNAP, we can treat middlebox contexts as state variables and transform FlowTags policies to SNAP programs. Thus, they can be easily composed with other SNAP policies. Next, we can fix the placement of middlebox state variables to the actual location of the middlebox in the network in SNAP's MILP. This way, SNAP's compiler can decide state placement and routing for SNAP's own policies while making sure that the paths between different middleboxes in the FlowTags policies exist in the network. Thus, steering happens using SNAP-generated tags. Middleboxes can still use tags from FlowTags to learn about flow's original information or the context of the previous middlebox.

Finally, we focus on *programming* networks but if verification is of interest in future work, one might adopt techniques such as RONO [24] to verify isolation prop-

erties in the presence of stateful middleboxes. In summary, interacting with existing middleboxes is no harder or easier in SNAP than it is in other *global* SDN languages, stateless or stateful, such as NetKAT [2] or Stateful NetKAT [18].

7.3 Extending SNAP

Sharding state variables. The MILP assigns each state variable to *one* physical switch to avoid the overhead of synchronizing multiple instances of the same variable. Still, distributing a state variable remains a valid option. For instance, the compiler can partition $s[inport]$ into k *disjoint* state variables, each storing s for one port. The MILP can decide placement and routing as before, this time with the option of distributing partitions of s with no concerns for synchronization.

Fault-Tolerance. SNAP's current prototype does not implement any particular fault tolerance mechanism in case a switch holding a state variable fails. Therefore, the state on the failed switch will be lost. However, this problem is not inherent or unique to SNAP and will happen in existing solutions with middleboxes too if the state of the middlebox is not replicated. Applying common fault tolerance techniques to switches with state to avoid state loss in case of failure can be an interesting direction for future work.

Modifying fields with state variables. An interesting extension to SNAP is allowing a packet field to be directly modified with the value of a state variable at a specific index: `f <- s[e]`. This action can be used in applications such as NATs and proxies, which can store connection mappings in state variables and modify packets accordingly as they fly by. Moreover, this action would enable SNAP programs to modify a field by the output of an arbitrary function on a set of packet fields, such as a hash function. Such a function is nothing but a fixed mapping between input header fields and output values. Thus, when analyzing the program, the compiler can treat these functions as fixed state variables with the function's input fields as index for the state variable and place them on switches with proper capabilities when distributing the program across the network. However, adding this action results in complicated dependencies between program statements, which is interesting to explore as future work.

Deep packet inspection (DPI). Several applications such as intrusion detection require searching the packet's payload for specific patterns. SNAP can be extended with an extra field called *content*, containing the packet's payload. Moreover, the semantics of tests on the content field can be extended to match on regular expressions. The compiler can also be modified to assign content tests to switches with DPI capabilities.

Resource constraints. SNAP's compiler optimizes state placement and routing for link utilization. However, other resources such as switch memory and processing power in terms of maximum number of complicated operations on packets (such as stateful updates,

increments, or decrements) may limit the possible computations on a switch. An interesting direction for future work would be to augment the SNAP compiler with the ability to optimize for these additional resources.

Cross-packet fields. Layer 4-7 fields are useful for classifying flows in stateful applications, but are often scattered across multiple physical packets. Middleboxes typically perform session reconstruction to extract these fields. Although SNAP language is agnostic to the chosen set of fields, the compiler currently supports fields stored *in the packet itself* and the state associated with them. However, it may be interesting to explore abstractions for expressing how multiple packets (e.g., in a session) can form "one big packet" and use its fields. The compiler can further place sub-programs that use cross-packet fields on devices that are capable of reconstructing the "one big packet".

Queue-based policies. SNAP currently has no notion of queues and therefore, cannot be used to express queue-based performance-oriented policies such as active queue management, queue-based load balancing, and packet scheduling. There is ongoing research on finding the right set of primitives for expressing such policies [34], which is largely orthogonal and complementary to SNAP's current goals.

8. RELATED WORK

Stateful languages. Stateful NetKAT [18], developed concurrently with SNAP, is a stateful language for "event-driven" network programming, which guarantees consistent update when transitioning between configurations in response to events. SNAP source language is richer and exponentially more compact than stateful NetKAT as it contains *multiple arrays* (as opposed to one) that can be indexed and updated by contents of *packet headers* (as opposed to constant integers only). Moreover, they place multiple copies of state at the edge, proactively generate rules for all configurations, and optimize for rule space, while we distribute state and optimize for congestion. Kinetic [17] provides a perflow state machine abstraction, and NetEgg [45] synthesizes stateful programs from user's examples. However, they both keep the state at the controller.

Compositional languages. NetCore [19], and other similar languages [20, 10, 2], have primitives for tests and modifications on packet fields as well as composition operators to combine programs. SNAP builds on these languages by adding primitives for stateful programming (§3). To capture the joint intent of two policies, sometimes the programmer needs to decompose them into their constituent pieces, and then reassemble them using ; and +. PGA [26] allows programmers to specify access control and service chain policies using graphs as the basic building block, and tackles this challenge by defining a new type of composition. However, PGA does not have linguistic primitives for stateful programming, such as those that read and write the con-

tents of global arrays. Thus, we view SNAP and PGA as complementary research projects, with each treating different aspects of the language design space.

Stateful switch-level mechanisms. FAST [21] and OpenState [4] propose flow-level state machines as a primitive for a *single* switch. SNAP offers a network-wide OBS programming model, with a compiler to distribute the programs across the network. Thus, although SNAP is exponentially more compact than a state machine in cases where state is indexed by contents of packet header fields, both FAST and OpenState can be used as a target for a subset of SNAP programs.

Optimizing placement and routing. Several projects have explored optimizing placement of middleboxes and/or routing traffic through them. These projects and SNAP share the mathematical problem of placement and routing on a graph. Merlin programs specify service chains as well as optimization objectives [39], and the compiler uses an MILP to choose paths for traffic with respect to specification. However, it does not decide the placement of service boxes itself. Rather, it chooses the paths to pass through the existing instances of the services in the physical network. Stratos [11] explores middlebox placement and distributing flows amongst them to minimize inter-rack traffic, and Slick [3] breaks middleboxes into fine-grained elements and distributes them across the network while minimizing congestion. However, they both have a separate algorithm for placement. In Stratos, placement results is used in an ILP to decide distribution of flows. Slick uses a virtual topology on the placed elements with heuristic link weights, and finds shortest paths between traffic endpoints.

9. CONCLUSION

In this paper, we introduced a stateful SDN programming model with a one-big-switch abstraction, persistent global arrays, and network transactions. We developed algorithms for analyzing and compiling programs, and distributing their state across the network. Based on these ideas, we prototyped and evaluated the SNAP language and compiler on numerous sample programs. We also explore several possible extensions to SNAP to support a wider range of stateful applications. Each of these extensions introduces new and interesting research problems to extend our language, compilation algorithms, and prototype.

Acknowledgments

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