

Sluice: A Network-Wide Model for Programmable Networks

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1 INTRODUCTION

The last several years have seen the emergence of programmable network devices including both programmable switching chips and programmable network interface cards (NICs). Along with the rise of x86-based packet processing for middleboxes and virtual switches, these trends point towards a future where the entire network will be programmable. The benefits of network programmability range from commercial use cases such as network virtualization implemented on Open vSwitch to more recent projects that implement packet scheduling, measurement, and application offload of niche applications on programmable switches.

While the benefits of programmability are clear, it is still difficult to program the network as a whole. Current programming languages target individual network devices, e.g., P4 for the Tofino programmable switching chip and the Netronome programmable NIC. However, at present, there is no unified programming model to express and implement general functionality at the level of an entire network, without having to individually program each network device.

Maple [39] was an early example of a network-wide programming model designed for OpenFlow switches. Maple automatically divided functionality between a stateless component running on switches and a stateful component running on the network’s controller. SNAP [40] is a more recent example of network-wide programming; unlike Maple, it additionally offloads stateful functionality to switches by leveraging stateful processing available in several programmable switches. However, both Maple and SNAP cannot express programmable-switch functionality that affects network performance at fine time scales, e.g., packet scheduling, congestion control, fine-grained measurement of microbursts, and load balancing. For these reasons we have developed Sluice, a programming model that takes a high-level specification of a network program and compiles it into runnable code that can be launched on the programmable devices of network. Sluice endows network operators with the ability to design and deploy large network programs for various functions such as scheduling, measurement, and application offloading. We demonstrate Sluice’s functionality and simplicity of use via two examples: traffic matrix generation for network analysis and a streaming join-filter operation.

2 SLUICE DESIGN

In the Sluice model, a network-wide program consists of high-level code *snippets* annotated by the operator to run on particular devices in a network. The code in each snippet is to be executed on packets arriving at its corresponding device. Snippets support a variety of operations : read-from/write-to packets; arithmetic using packet data, local variables, or stateful register arrays; control flow statements. To handle computation on custom packet headers not supported by default (ip/tcp/udp/eth), users may define packet declarations similar to C structs. An optional annotation in the declaration, the parser condition, automatically generates a header parser. This lets the user restrict snippets to operate on specific flows or IP address ranges. Sluice programs may also import device-specific variables/attributes for use in code snippets.

The compiler translates each snippet of a sluice program into a device-specific program. After initial parsing, a snippet is decomposed into a directed acyclic graph (DAG) that maps dependencies between variables in each snippet. This graph is then passed to the backend of the compiler that generates the corresponding P4 program for that device, for example bmv2 or Tofino ¹. Sluice automatically reduces the amount of boilerplate code needed to write an equivalent program in P4. For instance, the 9 line traffic matrix sluice program translates into over 200 lines of P4 code. Overall, Sluice provides the same functionality as P4 but makes it easier to program the data plane of an entire network

3 DEMONSTRATIONS

3.1 Traffic Matrix

Figure 2 displays the Mininet network topology and simulation components. Packets are sent over UDP from each host to all other hosts according to a poisson traffic model with mean inter-arrival time of 0.5 seconds. The codelet below is our Sluice program with a single snippet *traffic_example* that is launched on all switches of the network. To run the simulation, the user passes the Sluice program and network topology to the compiler. The compiler generates P4 code to run on each switch as well as control plane table entries for

¹Currently we only support bmv2 but plan to support more devices

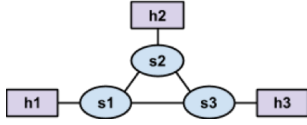


Figure 1: Topology For Traffic Matrix Demo

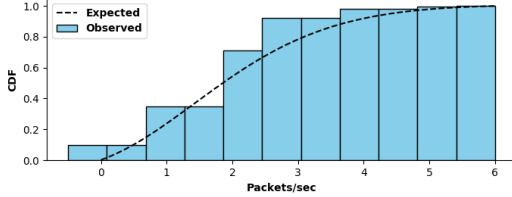


Figure 2: CDF of Packet Rate on link s1-s3

routing packets through the topology.

```
import device psa;

packet p: udp(srcPort:1234)
  nhops : bit <32>;

@ bmv2 : ;
snippet traffic_example()
  persistent cnt : bit <32>[10];
  cnt[psa.ingress_port] = cnt[psa.ingress_port] + 1;
  p.nhops = p.nhops + 1;
```

This demo shows how a simple Sluice program can be used to enable each switch to measure link usage for a specific flow of user-defined packets on UDP srcPort 1234. Each packet p contains a custom header $nhops$ that is incremented each time the packet enters a switch to inform the receiving host of the number of hops the packet took. Each switch maintains a stateful register counter cnt , indexed by switch ingress port, that tracks how many packets have entered through that ingress port. Aggregated over all switches, these counters data represent a matrix measuring each link's usage in the network at a given time. This matrix (residing on the whole network) is then queried once every second from the control plane to generate time-series plots of link-utilization for each link. Figure 3 displays the histogram of packet rates on link s1-s3 after collecting data for 3 minutes. The expected distribution of packet rates $poisson(\mu = 2 \text{ packets/sec})$ is also plotted to confirm the accuracy of the simulation.

3.2 Streaming Algorithms

This example demonstrates a simple join-filter operation between two streams. A stream is an unbounded table where

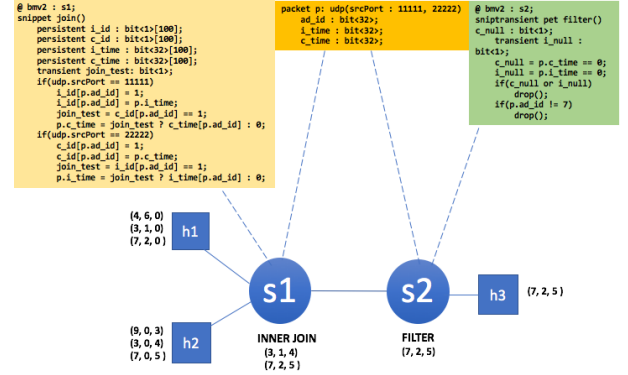


Figure 3: streaming example topology, data flow, and code placement on switches

a packet represents a tuple of data (ad_id , $impression_time$, $click_time$) enclosed in a custom header. The topology in Figure 4 describes the data flow and shows how an operator query runs on the switches of the network. Host 1 sends a stream of ad impressions while Host 2 sends a stream of ad clicks. The two streams are joined on the ad_id field at s1 and filtered on the ad_id field at s2 and the result is sent to h3.

4 FUTURE WORK

We plan to focus on using the dependency DAG to provide several optimizations. For example, it is possible that certain lines of code in a snippet cannot be run on the device annotated by the operator. Since programmable switching chips do not support floating point or complex string operations, code containing such features must be moved to the control plane or end host while at the same time, preserving the original program semantics intended by the operator. The issue of code movement is particularly interesting in the context of multi-tenant data centers. Here, each tenant manages their own virtual network, which is mapped to physical nodes at the data center. If each tenant wants to run their own network-wide program on their virtual topology, the data center operator will need to merge all these into one data plane implementation that runs on the entire physical network. We will continue to develop Sluice with the hope that it may virtualize the data plane, in the same way as network virtualization virtualized the control plane.

5 REFERENCES