

Millions of Little Minions: Using Packets for Low Latency Network Programming and Visibility (Extended Version)

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ABSTRACT

This paper presents a practical approach to rapidly introducing new dataplane functionality into networks: End-hosts embed tiny programs into packets to actively query and manipulate a network’s internal state. We show how this “tiny packet program” (TPP) interface gives end-hosts unprecedented visibility into network behavior, enabling them to work with the network to achieve a desired functionality. Our design leverages what each component does best: (a) switches forward and execute tiny packet programs (at most 5 instructions) in-band at line rate, and (b) end-hosts perform arbitrary (and easily updated) computation on network state. By implementing three different research proposals, we show that TPPs are *useful*. Using a hardware prototype on a NetFPGA, we show our design is *feasible* at a reasonable cost.

1 Introduction

Consider a large datacenter network with thousands of switches. Applications complain about poor performance due to high flow completion times for a small subset of their flows. As an operator, you realize this symptom could be due to congestion, either from competing cross traffic or poor routing decisions, or alternatively could be due to packet drops at failed links. In any case, your goal is to diagnose this issue quickly. Unfortunately, the extensive use of multi-path routing in today’s networks means one often cannot determine the exact path taken by every packet; hence it is quite difficult to triangulate problems to a single switch. Making matters worse, if congestion is intermittent, counters within the network will look “normal” at timescales of minutes or even seconds.

Such issues would be straightforward to debug if one could examine relevant network state such as switch ID, queue occupancy, input/output ports, port utilization, and matched forwarding rules *at the exact time each packet was forwarded*, so as to reconstruct what exactly transpired in the dataplane. In the above example, end-hosts could use state obtained from millions of successfully delivered packets to explicitly pinpoint network links that have high queue occu-

pancy (for congestion), or use switch and port IDs to verify that packets were correctly routed, or use path information to triangulate network links that cause packet drops due to link failures. In short, the ability to correlated network state to specific packets would be invaluable.

Can packets be instrumented to access and report on switch state? To date such state has been locked inside switches. This paper describes a *simple, programmable* interface that enables end-hosts to query switch memory (counters, forwarding table entries, etc.) from packets, directly in the dataplane. Specifically, a subset of packets carry in their header a tiny packet program (TPP), which consists of a few instructions that read, write, or perform simple, protocol-agnostic computation using switch memory.

A key observation in this paper is that having such programmable and *fast* access to network state benefits a broad class of network tasks—congestion control, measurement, troubleshooting, and verification—which we call *dataplane* tasks. We show how the TPP interface enables end-hosts to rapidly deploy new functionality by refactoring many network tasks into: (a) simple TPPs that execute on switches, and (b) expressive programs at end-hosts.

TPPs contrast to three approaches to introduce new dataplane functionality: (1) build custom hardware for each task, (2) build switches that can execute arbitrary code [36, 39], or (3) use FPGAs and network processors [29]. Each approach has its own drawbacks: Introducing new switch functionality can take many years; switch hardware has stringent performance requirements and cannot incur the penalty of executing arbitrary code; and FPGAs and network processors are simply too expensive at large scale [9]. Instead, we argue that if we could build new hardware to support just *one* simple interface such as the TPP, we can leverage end-hosts to implement *many* complex tasks at software-development timescales.

TPPs can be viewed as a particular, reasoned point within the spectrum of ideas in Active Networking [36, 39]. In many Active Networks formulations, routers execute arbitrary programs that actively control network behavior such as routing, packet compression, and (de-)duplication. By

Instruction	Meaning
LOAD, PUSH	Copy values from switch to packet
STORE, POP	Copy values from packet to switch
CSTORE	Conditionally store and execute subsequent operations
CEXEC	Conditionally execute the subsequent instructions

Table 1: The tasks we present in the paper require support only for the above instructions, whose operands will be clear when we discuss examples. Write instructions may be selectively disabled by the administrator.

contrast, TPP instructions are so simple they execute within the time to forward packets at line-rate. Just a handful of TPP instructions, shown in Table 1, providing access to the statistics in Table 2, proved sufficient to implement several previous research proposals.

1.1 Goals

Our main goal is to expose network state to end-hosts through the dataplane. To benefit dataplane tasks, any interface should satisfy the following requirements:

- **Speed:** A recent study shows evidence that switch CPUs are not powerful and are unable to handle more than a few hundred OpenFlow control messages/second [17]. Our experience is that such limitations stand in the way of a whole class of dataplane tasks as they operate at packet and round-trip timescales.
- **Packet-level consistency:** Switch state such as link queue occupancy and forwarding tables varies over time. Today, we lack any means of obtaining a consistent view of such state as it pertains to each packet traveling through the network.
- **Minimality and power:** To be worth the effort, any hardware design should be simple, be sufficiently expressive to enable a diverse class of useful tasks, and incur low-enough overhead to work at line rates.

This paper presents a specific TPP interface whose design is largely guided by the above requirements.

Non-Goals: It is worth noting that our goal is not to be flexible enough to implement any, and all dataplane network tasks. For instance, TPPs are not expressive enough to implement per-packet scheduling. Moreover, our design is for networks owned and operated by a single administrative entity (e.g., privately owned WANs and datacenters). We do not advocate exposing network state to untrusted end-hosts connected to the network, but we describe mechanisms to avoid executing untrusted TPPs (§4.3). Finally, a detailed design for inter-operability across devices from multiple vendors is beyond the scope of this paper, though we discuss one plausible approach (§8).

1.2 Summary of Results

Through both a software implementation and a NetFPGA prototype, this paper demonstrates that TPPs are both useful and feasible at line rate. Moreover, an analysis using recent

Statistics	Examples
Per-Switch	Switch ID, counters associated with the global L2 or L3 flow tables, flow table version number, timestamp.
Per-Port	Link utilization, bytes received, bytes dropped, bytes enqueued, application-specific registers.
Per-Queue	Bytes enqueued, bytes dropped.
Per-Packet	Packet’s input/output port, queue, matched flow entry, alternate routes for a packet.

Table 2: A non-exhaustive list of statistics stored in switches memory that TPPs can access when mapped to known memory locations. Many statistics are already tracked today but others, such as flow table version will have to be implemented. Some statistics are read-only (e.g. matched flow entry, bytes received), but others can be modified (e.g. packet’s output port). See OpenFlow 1.4 specification [32, Table 5] for a detailed list of available statistics.

data [9] suggests that TPP support within switch hardware can be realized at an acceptable cost.

Applications: We show the benefits of TPP by refactoring many recent research proposals using the TPP interface. These tasks broadly fall under the following three categories:

- **Congestion Control:** We show how end-hosts, by periodically querying network link utilization and queue sizes with TPP, can implement a rate-based congestion control algorithm (RCP) providing max-min fairness across flows. We furthermore show how the TPP interface enables fairness metrics beyond the max-min fairness for which RCP was originally designed (§2.2).
- **Network Troubleshooting:** TPPs give end-hosts detailed per-packet visibility into network state that can be used to implement a recently proposed troubleshooting platform called NetSight [16]. In particular, we walk through implementing and deploying ndb, a generalization of traceroute introduced by NetSight (§2.3).
- **Network Monitoring:** TPPs can be used in a straightforward way to do network monitoring, but we also show how to refactor new kinds of measurement tasks: For example, OpenSketch [42] proposed switch modifications to increase accuracy of five different measurement tasks while incurring low storage overhead. We show how to achieve similar functionality using network visibility offered by TPPs. In particular, we walk through using TPPs to count the number of unique source IP addresses that communicate over all network links in the core of the network (§2.5).
- **Network Control:** We also demonstrate how low-latency visibility offered by TPPs enables end-hosts to control how traffic is load balanced across network paths. We refactor CONGA [1], an in-network load-balancing mechanism implemented in Cisco’s new ASICs, between end-hosts and a network that supports only the TPP interface.

Hardware: To evaluate the feasibility of building a TPP-capable switch, we synthesized and built a four-port NetFPGA router (at 160MHz) with full TPP support, capable of switching minimum sized packets on each interface at

10Gb/s. We show the hardware and latency costs of adding TPP support are minimal on NetFPGA, and argue the same would hold of a real switch (§6). We find that the key to achieving high performance is restricting TPPs to a handful of instructions per packet (say five), as it ensures that any TPP executes within a fraction of the its transmission time.

Software: We also implemented the TPP interface in Open vSwitch [34], which we use to demonstrate research proposals and examples. Additionally, we present a software stack (§4) that enforces security and access control, handles TPP composition, and has a library of useful primitives to ease the path to deploying TPP applications.

The software and hardware implementations of TPP, scripts to run experiments and plots in this paper, and an extended version of this paper describing more TPP applications are all available online at <http://jvimal.github.io/tpp>.

2 Example Programs

We start our discussion using examples of dataplane tasks that can be implemented using TPPs, showcasing the utility of exposing network state to end-hosts directly in the dataplane. Each of these tasks typically requires new task-specific hardware changes; however, we show how each task can be refactored such that the network only implements TPPs, while delegating complex task-specific functionality to end-hosts. We will discuss the following tasks: (i) micro-burst detection, (ii) a rate-based congestion control algorithm, (iii) a network troubleshooting platform, (iv) a congestion aware, distributed, network load balancer, and (v) a low-overhead network measurement platform.

What is a TPP? A TPP is any Ethernet packet with a uniquely identifiable header that contains instructions, some additional space (packet memory), and an optional encapsulated Ethernet payload (e.g. IP packet). The TPP exclusively owns its packet memory, but also has access to shared memory on the switch (its SRAM and internal registers) through addresses. TPPs execute directly in the dataplane at every hop, and are forwarded just like other packets. TPPs use a very minimal instruction set listed in Table 1, and we refer the reader to Section 3 to understand the space overheads. We abuse terminology, and use TPPs to refer both to the programs and the packets that carry them.

We write TPPs in a pseudo-assembly-language with a segmented address space naming various registers, switch RAM, and packet memory. We write addresses using human-readable labels, such as `[Namespace:Statistic]` or `[Queue:QueueOccupancy]`. We posit that these addresses be known upfront at compile time. For example, the mnemonic `[Queue:QueueOccupancy]` could be refer to an address `0xb000` that stores the occupancy of a packet’s output queue at each switch.

2.1 Micro-burst Detection

Consider the problem of monitoring link queue occupancy within the network to diagnose short-lived congestion events

(or “micro-bursts”), which directly quantifies the impact of incast. In low-latency networks such as datacenters, queue occupancy changes rapidly at timescales of a few RTTs. Thus, observing and controlling such bursty traffic requires visibility at timescales orders of magnitude faster than the mechanisms such as SNMP or embedded web servers that we have today, which operate at tens of seconds at best. Moreover, even if the monitoring mechanism is fast, it is not clear which queues to monitor, as (i) the underlying routing could change, and (ii) switch hash functions that affect multipath routing are often proprietary and unknown.

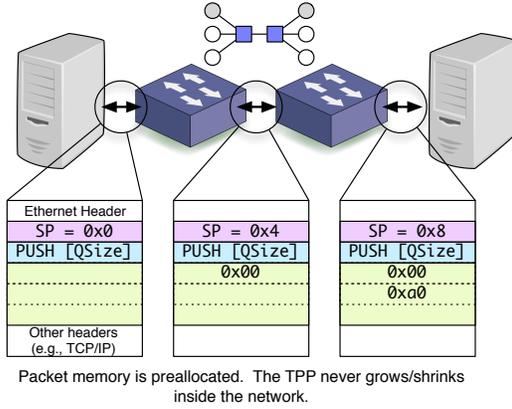
TPPs can provide fine-grained per-RTT, or even per-packet visibility into queue evolution inside the network. Today, switches already track per-port, per-queue occupancy for memory management. The instruction `PUSH [Queue:QueueOccupancy]` could be used to copy the queue register onto the packet. As the packet traverses each hop, the packet memory has snapshots of queue sizes at each hop. The queue sizes are useful in diagnosing micro-bursts, as they are not an average value. They are recorded when the packet traverses the switch. Figure 1a shows the state of a sample packet as it traverses a network. In the figure, SP is the stack pointer which points to the next offset inside the packet memory where new values may be pushed. Since the maximum number of hops is small within a datacenter (typically 5–7), the end-host preallocates enough packet memory to store queue sizes. Moreover, the end-host knows exactly how to interpret values in the packet to obtain a detailed breakdown of queueing latencies on all network hops.

This example illustrates how a low-latency, programmatic interface to access dataplane state can be used by software at end-hosts to measure dataplane behavior that is hard to observe in the control plane. Figure 1a shows a six-node dumbbell topology on Mininet [15], in which each node sends a small 10kB message to every other node in the topology. The total application-level offered load is 30% of the hosts’ network capacity (100Mb/s). We instrumented every packet with a TPP, and collected fully executed TPPs carrying network state at one host. Figure 1b shows the queue evolution of 6 queues inside the network obtained from every packet received at that host.

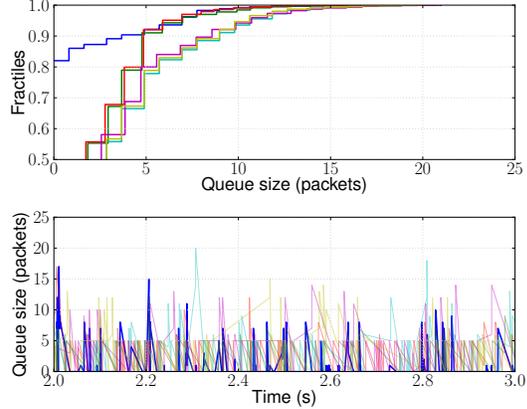
Overheads: The actual TPP consists of three instructions, one each to read the switch ID, the port number, and the queue size, each a 16 bit integer. If the diameter of the network is 5 hops, then each TPP adds only a 54 byte overhead to each packet: 12 bytes for the TPP header (see §3.4), 12 bytes for instructions, and 6×5 bytes to collect statistics at each hop.

2.2 Rate-based Congestion Control

While the previous example shows how TPPs can help observe latency spikes, we now show how such visibility can be used to *control* network congestion. Congestion control is arguably a dataplane task, and the literature has a number of ideas on designing better algorithms, many of which



(a) Visualizing the execution of a TPP as it is routed through the network.



(b) CDF and time series of queue occupancy on 6 queues in the network, obtained from every packet arriving at one host.

Figure 1: TPPs enable end-hosts to measure queue occupancy evolution at a packet granularity allowing them to detect microbursts, which are the spikes in the time series of queue occupancy (bottom of Figure 1b). Notice from the CDF (top) that one of the queues is empty for 80% of the time instants when packet arrives to the queue; a sampling method is likely to miss the bursts.

require switch support. However, TCP and its variants still remain the dominant congestion control algorithms. Many congestion control algorithms, such as XCP [23], FCP [14], RCP [11], etc. work by monitoring state that indicates congestion and adjusting flow rates every few RTTs.

We now show how end-hosts can use TPPs to deploy a new congestion control algorithm that enjoys many benefits of in-network algorithms, such as Rate Control Protocol (RCP) [11]. RCP is a congestion control algorithm that rapidly allocates link capacity to help flows finish quickly. An RCP router maintains one fair-share rate $R(t)$ per link (of capacity C , regardless of the number of flows), computed periodically (every T seconds) as follows:

$$R(t+T) = R(t) \left(1 - \frac{T}{d} \times \frac{a(y(t) - C) + b \frac{q(t)}{d}}{C} \right) \quad (1)$$

Here, $y(t)$ is the average ingress link utilization, $q(t)$ is the average queue size, d is the average round-trip time of flows traversing the link, and a and b are configurable parameters. Each router checks if its estimate of $R(t)$ is smaller than the flow's fair-share (indicated on each packet's header); if so, it replaces the flow's fair share header value with $R(t)$.

We now describe RCP*, an end-host implementation of RCP. The implementation consists of a rate limiter and a rate controller at end-hosts for every flow (since RCP operates at a per-flow granularity). The network control plane allocates two memory addresses per link (`Link:AppSpecific_0` and `Link:AppSpecific_1`) to store fair rates. Each flow's rate controller periodically (using the flow's packets, or using additional probe packets) queries and modifies network state in three phases.

Phase 1: Collect. Using the following TPP, the rate controller queries the network for the switch ID on each hop,

queue sizes, link utilization, and the link's fair share rate (and its version number), for all links along the path. The receiver simply echos a fully executed TPP back to the sender. The network updates link utilization counters every millisecond. If needed, end-hosts can measure them faster by querying for `[Link:RX-Bytes]`.

```
PUSH [Switch:SwitchID]
PUSH [Link:QueueSize]
PUSH [Link:RX-Utilization]
PUSH [Link:AppSpecific_0] # Version number
PUSH [Link:AppSpecific_1] # Rfair
```

Phase 2: Compute. In the second phase, each sender computes a fair share rate R_{link} for each link: Using the samples collected in phase 1, the rate controller computes the average queue sizes on each link along the path. Then, it computes a per-link rate R_{link} using the RCP control equation.

Phase 3: Update. In the last phase, the rate-controller of each flow asynchronously sends the following TPP to update the fair rates on all links. To ensure correctness due to concurrent updates, we use the `CSTORE` instruction:

```
CSTORE [Link:AppSpecific_0], \
      [Packet:Hop[0]], [Packet:Hop[1]]
STORE [Link:AppSpecific_1], [Packet:Hop[2]]
PacketMemory:
Hop1: V_1, V_1+1, R_new_1, (* 16 bits each*)
Hop2: V_2, V_2+1, R_new_2, ...
```

where V_i is the version number in the `AppSpecific_0` that the end-host used to derive an updated $R_{\text{new},i}$ for hop i , thus ensuring consistency. (`CSTORE dst,old,new` updates `dst` with `new` only if `dst` was `old`, ignoring the rest of the TPP

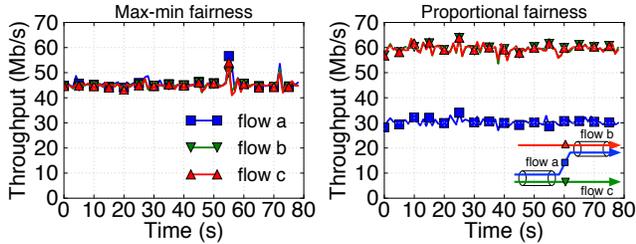


Figure 2: Allocations by max-min and proportional fairness variant of RCP on the traffic pattern shown inset on the right plot; each link has 100Mb/s capacity and all flows start at 1Mb/s at time 0.

otherwise.) Note that in the TPP, the version numbers and fair rates are read from packet memory at every hop.

Other allocations: Although RCP was originally designed to allocate bandwidth in a max-min fair manner among competing flows, Kelly et al. [25] showed how to tweak RCP to allocate bandwidth for a spectrum of fairness criteria— α -fairness parameterized by a real number $\alpha \geq 0$. α -fairness is achieved as follows: if R_i is the fair rate computed at the i -th link traversed by the flow (as per the RCP control equation 1), the flow sets its rate as

$$R = \left(\sum_i R_i^{-\alpha} \right)^{-1/\alpha} \quad (2)$$

The value $\alpha = 1$ corresponds to proportional fairness, and we can see that in the limit as $\alpha \rightarrow \infty$, $R = \min_i R_i$, which is consistent with the notion of max-min fairness. Observe that if the ASIC hardware had been designed for max-min version of RCP, it would have been difficult for end-hosts to achieve other useful notions of fairness. However, TPPs help defer the choice of fairness to deployment time, as the end-hosts can aggregate the per-link R_i according to equation 2 based on *one* chosen α . (We do not recommend flows with different α sharing the same links due to reasons in [38].)

Figure 2 shows the throughput of three flows for both max-min RCP* and proportional-fair RCP* in Mininet: Flow ‘a’ shares one link each with flows ‘b’ and ‘c’ (shown inset in the right plot). Flows are basically rate-limited UDP streams, where rates are determined using the control algorithm: Max-min fairness should allocate rates equally across flows, whereas proportional fairness should allocate $\frac{1}{3}$ of the link to the flow that traverses two links, and $\frac{2}{3}$ to the flows that traverse only one link.

Overheads: For the experiment in Figure 2, the bandwidth overhead imposed by TPP control packets was about 1.0–6.0% of the flows’ rate as we varied the number of long lived flows from 3 to 30 to 99 (averaged over 3 runs). In the same experiment, TCP had slightly lower overheads: 0.8–2.4%. The RCP* overhead is in the same range as TCP because each flow sends control packets roughly once every RTT. As the number of flows n increases, the average per-flow rate decreases as $1/n$, which causes the RTT of each flow to

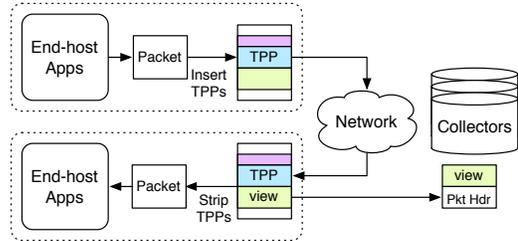


Figure 3: TPPs enable end-hosts to efficiently collect packet histories, which can then be used to implement four different troubleshooting applications described in [16].

increase (as the RTT is inversely proportional to flow rate). Therefore, the total overhead does not blow up.

Are writes absolutely necessary? RCP* is one of the few TPP applications that *writes* to network state. It is worth asking if this is absolutely necessary. We believe it is necessary for fast convergence since RCP relies on flows traversing a single bottleneck link agreeing on *one* shared rate, which is explicitly enforced in RCP. Alternatively, if rapid convergence isn’t critical, flows can converge to their fair rates in an AIMD fashion *without* writing to network state. In fact, XCP implements this AIMD approach, but experiments in [11] show that XCP converges more slowly than RCP.

2.3 Network Troubleshooting Framework

There has been recent interest in designing programmatic tools for troubleshooting networks; without doubt, dataplane visibility is central to a troubleshooter. For example, consider the task of verifying that network forwarding rules match the intent specified by the administrator [24, 26]. This task is hard as forwarding rules change constantly, and a network-wide ‘consistent’ update is not a trivial task [35]. Verification is further complicated by the fact that there can be a mismatch between the control plane’s view of routing state and the actual forwarding state in hardware (and such problems have shown up in a cloud provider’s production network [27]). Thus, verifying whether packets have been correctly forwarded requires help from the dataplane.

Recently, researchers have proposed a platform called NetSight [16]. NetSight introduced the notion of a ‘packet history,’ which is a record of the packet’s path through the network and the switch forwarding state applied to the packet. Using this construct, the authors show how to build four different network troubleshooting applications.

We first show how to efficiently capture packet histories that are central to the NetSight platform. NetSight works by interposing on the control channel between the controller and the network, stamping each flow entry with a unique version number, and modifying flow entries to create truncated copies of packet headers tagged with the version number (without affecting a packet’s normal forwarding) and additional metadata (e.g., the packet’s input/output ports). These truncated packet copies are reassembled by servers to reconstruct the packet history.

We can refactor the task of collecting packet histories by having a trusted agent at every end-host (§4) insert the TPP shown below on all (or a subset of) its packets. On receiving a TPP that has finished executing on all hops, the end-host gets an accurate view of the network forwarding state that affected the packet’s forwarding, without requiring the network to create additional packet copies.

```
PUSH [Switch:ID]
PUSH [PacketMetadata:MatchedEntryID]
PUSH [PacketMetadata:InputPort]
```

Once the end-host constructs a packet history, it is forwarded to collectors where they can be used in many ways. For instance, if the end-host stores the histories, we get the same functionality as `netshark`—a network-wide `tcpdump` distributed across servers. From the stored traces, an administrator can use any query language (e.g., SQL) to extract relevant packet histories, which gives the same functionality as the interactive network debugger `ndb`. Another application, `netwatch` simply uses the packet histories to verify whether network forwarding trace conforms to a policy specified by the control plane (e.g., isolation between tenants).

Overheads: The instruction overhead is 12 bytes/packet and 6 bytes of per-hop data. With a TPP header and space for 10 hops, this is 84 bytes/packet. If the average packet size is 1000 bytes, this is a 8.4% bandwidth overhead if we insert the TPP on every packet. If we enable it only for a subset of packets, the overhead will be correspondingly lower.

Caveats: Despite its benefits, there are drawbacks to using only TPPs, especially if the network transforms packets in erroneous or non-invertible ways. We can overcome dropped packets by sending packets that will be dropped to a collector (we describe how in §2.6). Some of these assumptions (trusting the dataplane to function correctly) are also made by `NetSight`, and we believe the advantages of TPPs outweigh its drawbacks. For instance, TPPs can collect more statistics, such as link utilization and queue occupancy, along with a packet’s forwarding history.

2.4 Distributed Load Balancing

We now show how end-hosts can use TPPs to probe for network congestion, and use this detailed visibility to load balance traffic in a distributed fashion. We demonstrate a simplified version of `CONGA` [1], which is an in-network scheme for traffic load balancing. `CONGA` strives to maximize network throughput and minimize the maximum network link utilization in a distributed fashion by having network switches maintain a table of path-level congestion metrics (e.g., quantized link utilization). Using this information, switches route small bursts of flows (“flowlets”) selfishly on the least loaded path. `CONGA` is optimized for datacenter network topologies; we refer the curious reader to [1] for more details.

`CONGA`’s design highlights two benefits relevant to our discussion. First, it uses explicit visibility by having

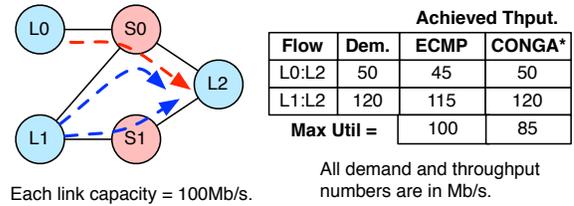


Figure 4: An example showing the benefits of congestion-aware load balancing: ECMP splits flow from L1 to L2 equally across the two paths resulting in suboptimal network utilization. CONGA*, an end-host refactoring of `CONGA` [1] is able to detect and reroute flows, achieving optimum in this example.

switches stamp quantized congestion information on packet headers. Second, load balancing decisions are made at round-trip timescales to rapidly detect and react to network congestion. Since TPPs also offer similar benefits, we show how we can refactor the load balancing task between end-hosts and the network, without requiring custom hardware (except, of course, to support TPPs).

First, we require the network to install multipath routes that end-hosts can select based on packet header values. This can be done in the slow-path by the control plane by programming a ‘group table’ available in many switches today for multipath routing [32, §5.6.1], which selects an output port by hashing on header fields (e.g., the VLAN tag). This allows end-hosts to select network paths simply by changing the VLAN ID.

Second, we need end-hosts to query for link utilization across various paths, by inserting the following TPP on a subset of packets destined to hosts within the datacenter:

```
PUSH [Link:ID]
PUSH [Link:TX-Utilization]
PUSH [Link:TX-Bytes]
```

We query for `Link:TX-Bytes` to measure small congestion events if the link utilization isn’t updated. The receiver echoes fully executed TPPs back to the sender to communicate the congestion. Note that the header of the echoed TPP also contains the path ID along with the link utilization on each link in the path.

Third, using information in the fully executed TPPs, end-hosts can build a table mapping ‘Path $i \rightarrow$ Congestion Metric (m_i),’ where m_i is either the maximum or sum of link utilization on each switch–switch network hop on path i . The authors of `CONGA` note that ‘sum’ is closer to optimal than ‘max’ in the worst-case scenario (adversarial); however `CONGA` used ‘max’ as it does not cause overflows when switches aggregate path-congestion. With TPPs, this is not an issue, and the choice can be deferred to deploy time.

And finally, end-hosts have full context about flows and flowlets, and therefore each end-host can select a flowlet’s path by setting the path tag appropriately on the flowlet’s packets.

Overheads: We implemented a proof-of-concept prototype (CONGA*) in software using UDP flows; Figure 4 reproduces an example from CONGA [1, Figure 4]. We configured switches S0 and S1 to select paths based on destination UDP port. The flow from L0 to L2 uses only one path, whereas the flow from L1 to L2 has two paths. The UDP agents at L0 and L1 query for link utilization and aggregate congestion metrics every millisecond for the two paths. With CONGA*, end-hosts can maximize network throughput meeting the demands for both flows, while simultaneously minimizing the maximum link utilization. In this example, the overhead introduced by TPP packets was minimal (< 1% of the total traffic).

Remark: Note that the functionality is *refactored* between the network and end-hosts; not all functionality resides completely at the end-hosts. The network implements TPP and multipath routing. The end-hosts merely select paths based on congestion completely in software.

2.5 Low-overhead Measurement

Since TPPs can read network state, it is straightforward to use them to monitor the network. However, we show how to implement non-trivial measurement tasks. In particular, OpenSketch [42] is a recently published measurement framework that makes the observation that many measurement tasks can be approximated accurately using probabilistic summary algorithms called “sketches,” which can in turn be compiled down to a three-stage pipeline where packets are hashed, filtered, and counted. The authors show how to combine these primitives to answer questions such as: (i) what is the number of unique IP addresses communicating across links? (ii) what is the flow size distribution across switches?

An example of a sketch is the bitmap sketch, which can count the number of unique elements in a stream as follows: hash the element (e.g. source IP address) to one of b bits and set it to 1. The estimate of the cardinality of the set is $b \times \ln(b/z)$, where z is the number of unset bits [13].

Many sketches require multiple hash functions in hardware operating at line-rate. This introduces a new ASIC functionality that is specific to sketches; thus, it is worth asking if the task can be refactored using the visibility offered by TPPs. Since end-hosts can readily implement many hash functions cheaply in software, the only piece of information they are missing is the packet’s routing context, which can be obtained by the following TPP:

```
PUSH [Switch:ID]
PUSH [PacketMetadata:OutputPort]
```

As example, consider a task where one wants to measure the number of unique destination IP addresses traversing the core switches in the network, and update these values every few (say 10) seconds. Each end-host in the network participates in this task by inserting the above TPP into its packets. Note that this TPP need not be inserted into all packets, but it

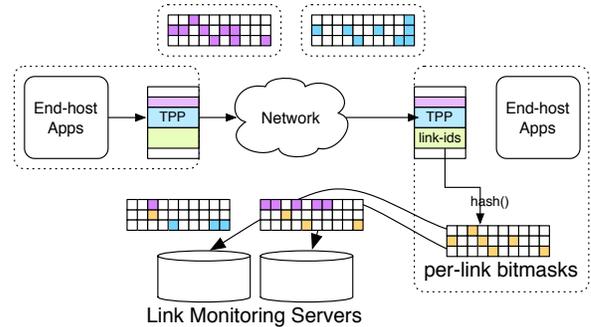


Figure 5: Refactoring the bitmap sketch [13] to estimate set cardinality.

should be inserted at least once for every destination IP address the host communicates with. The receiving end-host parses the fully executed TPP to retrieve the (switch,link) IDs from the packet and implements the sketch as follows:

```
index = hash(packet.ip.dest)
foreach (switch,link) in tpp:
    bitmask[switch][index] = 1
```

The sketch data-structures are distributed across the end-hosts in the network, but we can take advantage of the fact that the sketch operation (‘bit-set’) is commutative. Every 10 seconds, the end-hosts push those summary data structures that have changed since the last interval to a central, load-balanced, link monitoring service. The link monitoring service aggregates the bit-vectors to obtain the sketch data-structure for every link, obtaining the same result as one would obtain using OpenSketch. This refactoring allow end-hosts to retain flexibility in implementing other kinds of sketches.

Overheads: To implement the count-cardinality sketch, we only need one TPP per unique destination IP address. If we sample one out of every 10 packets to insert the measurement TPPs, we will incur less than 1% bandwidth overhead due to extra headers in packets. If we assume a $k = 64$ Fat-Tree datacenter network, there are 65536 core links, and 65536 servers. The sketch data-structure’s accuracy depends on the number of bits and the probability of collision [13]. If we use 1kbit memory per link, the total memory usage for all 65536 links is about 8MB/server.

2.6 Other possibilities

The above examples illustrate how a single TPP interface enables end-hosts to achieve many tasks. There are more tasks that we couldn’t cover in detail. In the interest of space, we refer the reader to the extended version of this paper for more details on some of the tasks below [31].

Measurement: Since TPPs can read network state, they can be used in a straightforward fashion for measuring any network statistic at rapid timescales. As TPPs operate in the

dataplane, they are in a unique position to expose path characteristics experienced by a specific packet that an end-host cares about.

Network verification: TPPs also help in verifying whether network devices meet certain requirements. For example, the path visibility offered by TPPs help accurately verify that route convergence times are within an acceptable value. This task can be challenging today, if we rely on end-to-end reachability as a way to measure convergence, because backup paths can still maintain end-to-end connectivity when routes change. Also, the explicit visibility eases fault localization [44].

Fast network updates: By allowing secure applications to write to a switch’s forwarding tables, network updates can be made very fast. This can reduce the time window of a transient state when network forwarding state hasn’t converged. For example, it is possible to add a new route to all switches along a path in half a round-trip time, as updating an IP forwarding table requires only 64 bits of information per-hop: 32 bit address and a 32 bit netmask per hop, tiny enough to fit inside a packet.

Wireless Networks: TPPs can also be used in wireless networks where access points can annotate end-host packets with rapidly changing state such as channel SNR. Low-latency access to such rapidly changing state is useful for network diagnosis, allows end-hosts to distinguish between congestive losses and losses due to poor channel quality, and query the bitrate that an AP selected for a particular packet.

3 Design of TPP-Capable Switches

In this section, we discuss the TPP instructions, addressing schemes, and the semantics of the TPP interface to a switch and what it means for a switch to be TPP-capable. Network switches have a variety of form factors and implementations; they could be implemented in software (e.g., Click, Open vSwitch), or in network processors (e.g., Net-FPGA), or as hardware ASICs. A switch might also be built hierarchically from multiple ASICs, as in ‘chassis’ based switches [21, Figure 3]. A TPP can be executed on each of these platforms. Thus, it is useful for a TPP-capable switch and the end-host to have a contract that preserves useful properties without imposing a performance penalty. We achieve this by constraining the instruction execution order and atomicity.

3.1 Background on a Switch Pipeline

We begin with an abstract model of a switch execution environment shown in Figure 6. The packet flows from input to output(s) through many pipelined modules. Once a packet arrives at an input port, the dataplane tags the packet with metadata (such as its ingress port number). Then, the packet passes through a parser that extracts fields from the packet and passes it further down the pipeline which consists of several match-action stages. This is also known as multiple match table model [9]. For example, one stage might

use the parsed fields to route the packet (using a combination of layer 2 MAC table, layer 3 longest-prefix match table, and a flexible TCAM table). Finally, any modifications to the packet are committed and the packet is queued in switch memory. Using metadata (such as the packet’s priority), the scheduler decides when it is time for the packet to be transmitted out of the egress port determined earlier in the pipeline. The egress stage also consists of a number of match-action stages.

3.2 TPP Semantics

The read/write instructions within a TPP access two distinct memory spaces: memory within the switch (switch memory), and a per-hop scratch space within the packet (packet memory). By all switch memory, we only mean memory at the stages traversed by a TPP, except the memory that stores packet contents. By all packet memory, we mean the TPP related fields in the packet. Now, we state our requirements for read/write instructions accessing the two memory spaces.

Switch memory: To expose statistics pertaining to a specific packet as it traverses the network, it is important for the instructions in the TPP to have access to the same values that are used to forward the packet. For read-only values, this requirement means that reads by a TPP to a single memory location must necessarily be atomic and after all writes by the forwarding logic to the *same* memory location. For example, if a TPP accesses the memory that holds the output port of a packet, it must return the same port that the forwarding logic determines, and no other value. This is what we mean by a “packet-consistent” view of network state.

For read-write memory addresses, it is useful if instructions within the TPP were executed in the order specified by the TPP to a given location *after* any modifications by the switch forwarding logic. Thus, writes by a TPP supersede those performed by forwarding logic.

Packet memory: Since instructions can read from and write to packet memory using PUSH and POP, writes to packet memory must take effect sequentially in the order specified by the TPP. This guarantees that if a TPP pushes values at memory locations X, Y, and Z onto packet memory, the end-host sees the values in the packet in the same order. This does not require that reads to X, Y, and Z be issued in the same order.

3.3 TPP Execution Model

TPPs are executed in the dataplane pipeline. TPPs are required to fit exactly within an MTU to avoid having the ASIC deal with fragmentation issues. This is not a big limitation, as end-hosts can split a complex task into multiple smaller TPPs if a single packet has insufficient memory to query all the required statistics. By default, a TPP executes at every hop, and instructions are not executed if they access memory that doesn’t exist. This ensures the TPP fails gracefully.

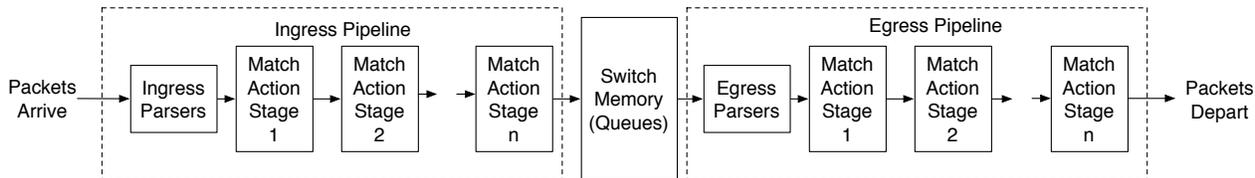


Figure 6: A simplified block diagram of the dataplane pipeline in a switch ASIC. Packets arrive at the ingress, and pass through multiple modules. The scheduler then forwards packets (that are not dropped) to the output ports computed at earlier stages in the pipeline.

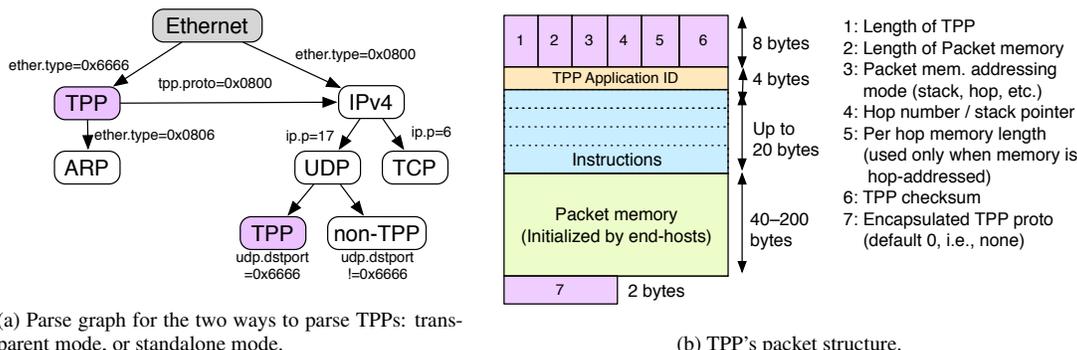


Figure 7: The parse graph and structure of a TPP. We chose 0x6666 as the ethertype and source UDP port that uniquely identifies a TPP. With a programmable switch parser, this choice can be reprogrammed at any time.

Furthermore, the platform is free to reorder reads and writes so they execute in any order. However, if the programmer needs guarantees on ordering instructions due to data hazards (e.g., for CEXEC, CSTORE), they must ensure the TPP accesses memory in the pipeline order. For a vast majority of use cases, we argue this restriction is not severe. From a practical standpoint, this requirement ensures that the switch pipeline remains feed-forward as it is today in a majority of switches.

3.3.1 Unified Memory-Mapped IO

A TPP has access to any statistic computed by the switch that is addressable. The statistics can be broadly namespaced into per-switch (i.e., global), per-port, per-queue and per-packet. Table 2 shows example statistics in each of these namespaces. These statistics may be scattered across different stages in the pipeline, but TPPs access them via a unified address space. For instance, a switch keeps metadata such as input port, the selected route, etc. for every packet that can be made addressable. These address mappings are known upfront to the TPP compiler that converts mnemonics such as [PacketMetadata:InputPort] into virtual addresses.

3.3.2 Addressing Packet Memory

Memory is managed using a stack pointer and a PUSH instruction that appends values to preallocated packet memory. TPPs also support a hop addressing scheme, similar to the the base:offset x86-addressing mode. Here, base:offset refers to the word at location $base * hop_size + offset$. Thus, if hop-size is 16 bytes, the instruction “LOAD [Switch:SwitchID], [Packet:hop[1]]” will copy the switch ID into

PacketMemory[1] on the first hop, PacketMemory[17] on the second hop, etc. The offset is part of the instruction; the base value (hop number) and per-hop memory size values are in the TPP header. To simplify memory management in the dataplane, the end-host must preallocate enough space in the TPP to hold per-hop data structures.

3.3.3 Synchronization Instructions

Besides read and write, a useful instruction in a concurrent programming environment is an atomic update instruction, such as a conditional store CSTORE, conditioned on a memory location matching a specified value, halting subsequent instructions in the TPP if the update fails. That is, CSTORE [X], [Packet:hop[Pre]], [Packet:hop[Post]] works as follows:

```

succeeded = False
if (value at X == value at Packet:hop[Pre]) {
    value at X = value at Packet:hop[Post]
    succeeded = True
}
value at Packet:hop[Pre] = value at X;
if (succeeded) {
    allow subsequent instructions to execute
}

```

By having CSTORE return the value of X, an end-host can infer if the instruction succeeded. Notice that the second and third operands are read from a unique location at every hop. This is needed to ensure correct semantics when the switch overwrites the value at the second operand.

In a similar vein, we found a conditional execute (CEXEC) instruction useful; for example, it may be desirable to ex-

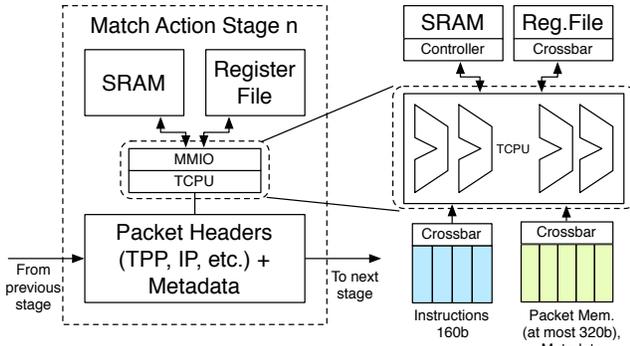


Figure 8: At every stage, the TCPU has execution units that can access only local memory and registers, as well as packet metadata.

ecute a network task only on one switch, or on a subset of switches (say all the top of rack switches in a datacenter). The conditional execute instruction specifies a memory address, a 32-bit mask, and a 32-bit value (specified in the packet hop), which instructs the switch to execute all subsequent instructions only when $(\text{switch_value} \& \text{mask}) == \text{value}$. All instructions that follow a failed CEXEC check will not be executed.

3.4 Parsing: TPP Packet Format

As noted in §2, a TPP is any Ethernet frame from which we can uniquely identify a TPP header, the instructions, packet memory, and an optional payload. This allows end-hosts to use TPPs in two ways: (i) piggy-back TPPs on any existing packet by encapsulating the packet within a TPP of ethernet type 0x6666, or (ii) embed a TPP into an otherwise normal UDP packet destined for port 0x6666, which is a special port number usurped by TPP-enabled routers.

Figure 7a shows the two parse graphs depicting the two ways in which our prototype uses TPPs. A parse graph depicts a state machine for a packet parser, in which the nodes denote protocols and edges denote state transitions when field values match. We use the same convention as in [9] to show the two ways in which we can parse TPPs.

3.5 Putting it together: the TCPU

TPPs execute on a tiny processor, which we call the TCPU. A simple way to implement the TCPU is by having a RISC-like processor at the end of the ingress match-action stages as we described in our earlier position paper [22, Figure 5]. This simple approach could be practical for software, or low-speed hardware switches, but might be impractical in high-speed hardware switches as memory in an ASIC is often distributed across modules. The wiring complexity to provide read and write paths from each module to the TCPU becomes prohibitively expensive within an ASIC, and is simply infeasible across line-cards in a chassis switch.

We overcome this limitation in two ways. First, our execution model permits reordering reads and writes across different ASIC memory locations. Second, end-hosts can stati-

cally analyze a desired TPP and split it into smaller TPPs if one TPP is insufficient. For instance, if an end-host requires link utilization on all links at all switches a packet traverses, it can stage the following sequence of TPPs: (i) send one TPP to collect switch ID and link utilizations on links traversed by the packet, and (ii) send a new TPP to each switch link on the switches traversed by TPP 1 to collect the remaining statistics. To summarize:

- Loads and stores in a single packet can be executed in any order, by having end-hosts ensure there are no write-after-write, or read-after-write conflicts.
- The operands for conditional instructions, such as CSTORE and CEXEC, are available before, or at the stages where the subsequent instructions execute; CEXEC can execute when all its operands are available.

By allowing instructions to be executed out of order, we can distribute the single logical TCPU on an ASIC by replicating its functionality at every stage. Each stage has one execution unit for every instruction in the packet, a crossbar to connect the execution units to all registers local to the stage and packet memory, and access to the stage’s local memory read/write port. From the decoded instructions, the stage can execute all instructions local to the stage, and once all memory accesses have completed, the packet leaves the stage.

Replicating execution units might seem expensive, but the majority of logic area in an ASIC is due to the large memories (for packet buffers, counters, etc.), so the cost of execution units is not prohibitive [9]. Figure 8 shows the TCPU if we zoom into one of the match-action stages.

Serializing PUSH/POP instructions: Finally, there are many techniques to ensure the effect of PUSH and POP instructions appear if they executed in order. Since the packet memory addresses accessed by PUSH/POP instructions are known immediately when they are parsed, they can be converted to equivalent LOAD/STOREs that can then be executed out of order. For example, consider the following TPP:

```
PUSH [PacketMetadata:OutputPort]
PUSH [PacketMetadata:InputPort]
PUSH [Stage1:Reg1]
POP [Stage3:Reg3]
```

After parsing the instructions, they can be converted to the following TPP which is equivalent to the above TPP:

```
LOAD [PacketMetadata:OutputPort], [Packet:Hop[0]]
LOAD [PacketMetadata:InputPort], [Packet:Hop[1]]
LOAD [Stage1:Reg1], [Packet:Hop[2]]
STORE [Stage3:Reg3], [Packet:Hop[2]]
```

Now, the TPP loads the values stored in two registers to the packet memory addressed in the hop addressing format. Note that the packet’s output port is not known until the packet is routed, i.e., at the end of the ingress stage. The execution proceeds as follows:

4.3 Security considerations

There is a great deal of software generating network traffic in any datacenter, and most of it should not be trusted to generate arbitrary TPPs. After all, TPPs can read and write a variety of switch state and affect packet routing. This raises the questions of how to restrict software from generating TPPs, but also how to provide some of the benefits of TPPs to software that is not completely trusted. We now discuss possible mechanisms to enforce such restrictions, under the assumption that switches are trusted, and there is a trusted software layer at end hosts such as a hypervisor.

Fortunately, restricting TPPs is relatively simple, because it boils down to packet filtering, which is already widely deployed. Just as untrusted software should not be able to spoof IP addresses or VLAN IDs, it should not be able to originate TPPs. Enforcing this restriction is as easy as filtering based on protocol and port numbers, which can be done either at all ingress switch ports or hypervisors.

In many settings, read-only access to most switch state is harmless. (A big exception is the contents of other buffered packets, to which TPPs do not provide access anyway.) Fortunately, TPPs are relatively amenable to static analysis, particularly since a TPP contains at most five instructions. Hence the hypervisor could be configured to drop any TPPs with write instructions (or write instructions to some subset of switch state). Alternatively, one could imagine the hypervisor implementing higher-level services (such as network visibility) using TPPs and expose them to untrusted software through a restricted API.

At a high level, a compromised hypervisor sending malicious TPPs is as bad as a compromised SDN controller. The difference is that hypervisors are typically spread throughout the datacenter on every machine and may present a larger attack surface than SDN controllers. Hence, for defense in depth, the control plane needs the ability to disable write instructions (STORE, CSTORE) entirely. A majority of the tasks we presented required only read access to network state.

4.4 TPP Executor

Although the default way of executing TPP is to execute at all hops from source to destination, we have built a ‘TPP Executor’ library that abstracts away common ways in which TPPs can be (i) executed reliably, despite TPPs being dropped in the network, (ii) targeted at one switch, without incurring a full round-trip from one end-host to another, (iii) executed in a scatter-gather fashion across a subset of switches, and many more. The goal of this section is to show how complex operations can be built from a simple set of primitives.

Reliable execution: Since TPPs are forwarded just like regular packets, they can be dropped if there is congestion. The helper function abstracts away retrying (for a maximum number of times) before giving up. Applications can use this functionality for idempotent operations (e.g., loads and conditional stores). We can make stores idempotent by first

reading the value and conditioning on the value for subsequent retries.

Targeted execution: TPPs can be crafted so that they execute only at one specific switch. This helper function wraps a TPP with a CEXEC instruction conditioned on the switch ID matching the specified value. The end-host agent creates a UDP packet and sends it to the switch IP (obtained from the network control plane).

As we noted in §3, a switch might have multiple pipelines connecting a set of interfaces. Since interfaces typically have their own IP addresses, the end-host can send a TPP addressed to a specific interface which would result in the TPP being routed through a pipeline. Furthermore, switches can be configured to reflected back to the source, using the following execution pattern.

Reflective TPP: Consider an example where a server wants to monitor congestion at the top-of-rack switch it is connected to. Sending a TPP from the server to a destination and back incurs a full round-trip time. To achieve even quicker response, the server can program the switch to reflect specially marked TPPs (in its header) back to the source address. With programmable protocol parsing [7, §4], switches can swap the source and destination IP addresses inside the TPP so the packet never leaves the rack.

Scatter gather: Some monitoring applications collect statistics from a number of switches, so we provide an executor that implements scatter-gather with retries. The application specifies the list of switches to execute a TPP, and the executor library takes care of creating the necessary TPPs, and masks any failures. The control plane can minimize the number of packets sent from the application to the switches, by constructing a multicast tree between the sender and monitored switches (in the slow path).

Large TPPs: If the statistics collected do not fit into a packet, either because the number of hops is large, or the number of statistics collected per-hop is large, then the executor automatically splits the TPP into smaller TPPs. The smaller TPPs use the CEXEC instruction, conditioned on the hop number on the packet’s TPP header, to execute on a specific range of hops.

4.5 Deploying a TPP Application

Applications that use TPPs can be deployed in two ways: (i) standalone, and (ii) piggy-backed. Standalone apps use the raw TPP interface, and have their own deployment strategy, because of their unique requirements. For example, RCP is a highly tuned implementation within the OS kernel at every end-host.

Piggy-backed applications are intrusive in the sense that they attach TPPs to packets that flow through the network. There is a common pattern to such applications (e.g., network troubleshooting and monitoring), which is abstracted away to make them easy to deploy. The programmer specifies the following:

- **Filter:** An `iptables` packet filter that specifies the subset of traffic through the network to which TPPs should be attached. The filter also includes the sampling frequency and the application’s priority.
- **TPP:** A compiled TPP (a string of bytes) that should be attached with packets.
- **Aggregator:** The aggregator is a per-node application that receives the application-specific, fully executed TPPs, and does post-processing. For example, the OpenSketch monitoring application implements hash functions and summary data-structures.
- **Collector:** A software service that collects summaries from the aggregators spawned across the cluster. The programmer specifies the service’s virtual IP address; packets sent to this IP are load balanced across a replicated collector instances.

Once the programmer specifies the above inputs, a provisioning agent creates a new application ID and verifies permissions by examining the TPP. The provisioning agent also spawns the aggregator and collector which receive fully executed TPPs and does application-specific processing. Finally, the provisioning agent configures appropriate end-host dataplane agents by invoking the `add_tpp` API call, completing the application setup. Once the application is set up, the collector will start receiving packets from its aggregator agents on deployed hosts.

5 Implementation

We have implemented both hardware and software support needed for TCPU: the distributed TCPU on the 10Gb/s NetFPGA platform, and a software TCPU for the Open vSwitch Linux kernel module. The NetFPGA hardware prototype has a four-stage pipeline at each port, with 64 kbit block RAM and 8 registers at each stage (i.e. a total of 1Mbit RAM and 128 registers). We were able to synthesize the hardware modules at 160 MHz, capable of switching minimum sized (64Byte) packets at a 40Gb/s total data rate.

The end-host stack is a relatively straightforward implementation: We have implemented the TPP-CP, and the TPP executor (with support only for the reliable and scatter-gather execution pattern) as Python programs running in userspace. The software dataplane is a kernel module that acts as a shim between the network stack and the underlying network device, where it can gain access to all network packets on the transmit and receive path. For filtering packets to attach TPPs, we use `iptables` to classify packets and tag them with a TPP number, and the dataplane inserts the appropriate TPP by modifying the packet in place.

6 Evaluation

In §2 we have already seen how TPPs enable many dataplane applications. We now delve into targeted benchmarks of the performance of each component in the hardware and software stack.

Task	NetFPGA	ASICs
Parsing	< 1 cycle	1 cycle
Memory access	1 cycle	2–5 cycles
Instr. Exec.: CSTORE	1 cycle	10 cycles
Instr. Exec.: (the rest)	< 1 cycle	1 cycle
Packet rewrite	< 1 cycle	1 cycle
Total per-stage	2–3 cycles	50–100 cycles†

Table 3: Summary of hardware latency costs. †The ASIC’s per-stage cost is estimated from the total end-to-end latency (200–500ns) and dividing it by the number of stages (typically 4–5). This does not include packetization latency, which is another ~50ns for a 64Byte packet at 10Gb/s.

6.1 Hardware

The cost of each instruction is dominated by the memory access latency. Instructions that only access registers complete in less than 1 cycle. On the NetFPGA, we use a single-port 128-bit wide block RAM that has a read (or write) latency of 1 cycle. We measured the total per-stage latency by sending a hundreds of 4 instruction TPP reading the clock from every stage, and found that the total per-stage latency was exactly 2 cycles: thus, parsing, execution, and packet rewrite all complete within a cycle, except for CSTORE, which takes 1 cycle to execute (excluding the time for accessing operands from memory).

The latency cost is different in a real switch: From personal communication with multiple ASIC designers [8, 10], we learned that 1GHz ASIC chips in the market typically use single-port SRAMs 32–128bits wide, and have a 2–5 cycle latency for every operation (read/write). This means that in the worst case, each load/store instruction adds a 5 cycle latency, and a CSTORE adds 10 cycles. Thus, in the worst case, if every instruction is a CSTORE, a TPP can add a maximum of 50ns latency to the pipeline; to avoid losing throughput due to pipeline stalls, we can add 50ns worth of buffering (at 1Tb/s, this is 6.25kB for the entire switch). However, the real cost is likely to be smaller because the ASIC already accesses memory locations that are likely to be accessed by the TPP that is being executed: For instance, the ASIC always looks up the flow entry, and updates queue sizes for memory accounting, so those values needn’t be read twice.

Though switch latency costs are different from that of the NetFPGA, they do not significantly impact packet processing latency, as in a typical workload, queueing and propagation delays dominate end-to-end latency and are orders of magnitude larger. Even within a switch, the unloaded ingress-egress latency for a commercial ASIC is about 500ns per packet [3]. The lowest-latency ASICs are in the range of about 200ns per packet [19]. Thus, the extra 50ns worst-case cost per packet adds at most 10–25% extra latency to the packet. Table 3 summarizes the latency costs.

Die Area: The NetFPGA costs are summarized in Table 4. Compared to the single-stage reference router, the costs are within 30.1% in terms of the number of gates. However, gate counts by themselves do not account for the total area cost, as logic only accounts for a small fraction of the total

Resource	Router	+TCPU	%-extra
Slices	26.8K	5.8K	21.6%
Slice registers	64.7K	14.0K	21.6%
LUTs	69.1K	20.8K	30.1%
LUT-flip flop pairs	88.8K	21.8K	24.5%

Table 4: Hardware cost of TPP modules at 4 pipelines in the NetFPGA (4 outputs, excluding the DMA pipeline).

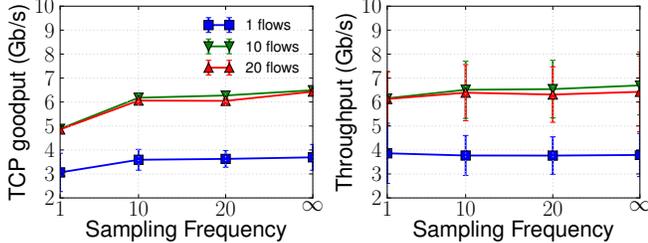


Figure 10: Maximum attainable application-level and network throughput with a 260 byte TPPs inserted on a fraction of packets (1500Byte MTU and 1240Byte MSS). A sampling frequency of ∞ depicts the baseline performance as no TPPs are installed. Error bars denote the standard deviation.

area that is dominated by memory. To assess the area cost for a real switch, we use data from Bosshart et al. [9]. In their paper, the authors note that the extra area for a total of 7000 processing units—which support instructions that are similar to the TCPU—distributed across all match-action stages, accounts for less than 7% of the ASIC area [9, §5.4]. We only need $5 \times 64 = 320$ TCPU, one per instruction per stage in the ingress/egress pipelines; therefore, the area costs are not substantial (0.32%).

6.2 End-host Stack

The critical component in the end-host stack is the dataplane. In the transmit side, the dataplane processes every packet, matches against a list of filters, and attaches TPPs. We use a 4-core Intel core i7 machine running Linux 3.12.6.

Figure 10 shows the baseline throughput of a single TCP flow, without segmentation offloads, across a virtual ethernet link, which was able to push about 4Gb/s traffic with one TCP flow, and about 6.5Gb/s of traffic with 20 flows. After adding TPPs, the throughput of the TCP flow reduces, depending on the (uniform random) sampling frequency. If the sampling frequency is infinite, none of the packets have TPPs, which denotes the best possible performance in our setup. As we can see, the network throughput doesn’t suffer much, which shows that the CPU overhead to add/remove TPPs is minimal. However, application throughput reduces proportionally, due to header overheads. Table 5 shows the

Match	# Rules				
	0	1	10	100	1000
First	8.8	8.7	8.6	7.8	3.6
Last	8.8	8.7	8.6	7.7	3.6
All	8.8	8.7	8.3	6.7	1.4

Table 5: Maximum attainable network throughput in Gb/s with varying number of filters (1500Byte MTU). The numbers are the average of 5 runs.

impact on the number of filters in the dataplane, and its effect on network throughput, under three different scenarios: (i) ‘first’ means we create flows that always match the first rule, (ii) ‘last’ means flows always match the last rule, and (iii) ‘all’ means there is at least one flow that matches each rule. In ‘first’ and ‘last,’ there are 10 TCP flows. In ‘all,’ there are as many flows as there are number of rules (with at least 10 flows). Each rule matches on a TCP destination port. As we can see, there is little loss in throughput up to 10 rules. With more rules, throughput does drop, but there is no difference between matching on the first (best case) and last rule (worst case) in the filter chain. With 1000 flows, other overheads (context switches) result in much lower throughput.

7 Limitations

Though TPPs help in a wide variety of tasks that were discussed in §2, they are not a panacea to implement any arbitrary functionality due to two reasons: (i) the restricted instruction set, and (ii) restricted programming model in which end-hosts initiate tasks. As we have not presented a formal theory of “network tasks,” the classification below is neither complete nor mutually exclusive; it is only meant to be an illustration.

Tasks that require per-packet computation: The read and write instructions in TPPs limit end-hosts to high throughput network *updates*, but not arbitrary network computation. As an example, consider the task of implementing an active queue management scheme such as Stochastic Fair Queuing, static priority, dynamic priority queueing (e.g. pFabric [2]), and fair queueing. These tasks require fine-grained control over per-packet transmit and drop schedules, which is better realized using dedicated hardware or FPGAs [37]. In a similar vein, TPPs are not expressive enough to scan packets for specific signatures (e.g., payload analysis using deep packet inspection). Such tasks are better served by other approaches (e.g., middlebox software, or custom packet processors).

Tasks that are event-driven: In the examples we discussed, all TPPs originate at end-hosts. This limits end-hosts from implementing tasks that require precisely timed notifications whenever there is some state change within the network. For instance, TPPs by themselves cannot be used to implement flow control mechanisms (e.g., priority flow control, or PFC [18]), or reactive congestion notifications such as Quantized Congestion Notification [33] and FastLane [43]. Such tasks require the network to send special packets when the queue occupancy reaches a certain threshold. However, this isn’t a show-stopper for TPPs, as end-hosts can proactively inject TPPs on a subset of packets and be notified quickly of network congestion.

8 Discussion

In §2, we showed how TPPs enable end-hosts to access network state with low-latency, which can then act on this state

to achieve a certain functionality. This is attractive as it enables interesting functionality to be deployed at software-development timescales. We now discuss a number of important concerns that we haven't covered.

Handling Device Heterogeneity: There are two issues here: instruction encoding, and statistics addressing. First, instructions are unlikely to be implemented in an ASIC as hardwired logic, but using microcodes, adding a layer of indirection for platform specific designs. Second, we recommend having two address spaces: (i) a standardized address space where a majority of the important statistics are preloaded at known locations, such as those identified by the OpenFlow standard [32], and (ii) a platform-specific address space through which additional statistics, specific to vendors and switch generations can be accessed. For dealing with multiple vendors, TPPs can support an indirect addressing scheme, so that the compiler can preload packet memory with platform specific addresses. For example, to load queue sizes from a Broadcom ASIC at hop 1, and an Intel ASIC at hop 2, the compiler generates the TPP below, loading the word from 0xff00 for Broadcom, and 0xfe00 for Intel, obtained out-of-band. For safety, the entire TPP is wrapped around a CEXEC as follows:

```
CEXEC [Switch:VendorID], [Packet:Hop[0]]
LOAD [[Packet:Hop[1]], [Packet:Hop[1]]
PacketMemory:
Hop1: $BroadcomVersionID, 0xff00 (* overwritten *)
Hop2: $IntelVersionID, 0xfe00
```

The TPP compiler can query the ASIC vendor IDs from time to time and change the addresses if the devices at a particular hop suddenly change. However, indirect addressing limits the extent to which a TPP can be statically analyzed.

MTU issues: Piggy-backed TPPs are attached to packets at the edge of a network (end-host or a border router). Thus, if the incoming packet is already at the MTU size, there would be no room to add a TPP. This is fortunately not a big issue, as many switches support MTUs up to 9000 bytes. This is already being done today in overlay networks to add headers for network virtualization [1].

Offloading more functionality to the network: Many network tasks we present in this paper requires cooperation from all end-hosts to achieve a task. It is worth asking if including every end-host is worth the trouble. We argue that it is: Many tasks such as congestion control (e.g. TCP), and monitoring (e.g. SNAP [41]), are already implemented in a way where every server takes part in the task. Network functionality can have a more informed path to a hardware implementation starting with the end-hosts.

Active networks and end-to-end arguments: Active Networks was criticized for its complexity and lack of 'killer applications,' but the end-to-end principle does not completely rule it out [6]. The networks we build are larger, more complex, and more critical than ever, as they form the backbone

of all compute within and between datacenters. To manage a network at large scale, applications must have an unprecedented visibility into the dataplane at timescales orders of magnitude faster than with the best tools we have today. We do not claim that TPPs are "novel," as they fall under the broad category of active networks. However, we have ruthlessly tried to keep the interface between the end-hosts and dataplane to the bare minimum to make interesting applications feasible. We believe TPPs strike a delicate balance between what is possible in hardware at line rate, and sufficient flexibility that end-hosts can implement dataplane tasks.

9 Related Work

TPPs represent a point in the broad design space of programmable networks, ranging from essentially arbitrary in-band programs as formulated by Active Network proposals [36, 39], to switch-centric programmable dataplane pipelines [4, 9, 20, 28], to controller-centric out-of-band proposals such as OpenFlow [30] and Simple Network Management Protocol (SNMP). We do not claim that the TPP approach is a fundamentally novel idea, as it is a specific realization of Active Networks. However, we have been ruthless in simplifying the interface between the end-hosts and switches to a bare minimum. We believe TPPs strike a delicate balance between what is possible in switch hardware at line rate, and what is sufficiently expressive for end-hosts to perform a variety of useful tasks.

TPPs superficially resemble Sprocket, the assembly language in Smart Packets [36]. However, Sprocket represents a far more expressive point in the design space. It allows loops and larger programs that would be hard to realize in hardware at line rate. By contrast, a TPP is a straight-line program whose execution latency is deterministic, small, and known at compile time. TPPs fully execute on the fast-path (i.e., router ASIC), whereas Sprocket exercises the slow-path (router CPU), which has orders of magnitude lower bandwidth. TPPs also resemble the read/write in-band control mechanism for ATM networks as described in a patent [5]; however, we also focus extensively on how to refactor useful dataplane tasks, and a security policy to safeguard the network against malicious TPPs. Wolf et al. [40] focus on designing a high performance Active Network router that supports general purpose instructions. It is unclear whether their model allows end-hosts to obtain a consistent view of network state. Moreover, it is unlikely that ASICs can take on general purpose computations at today's switching capacities at a reasonable cost. Furthermore, out-of-band control mechanisms such as OpenFlow and Simple Network Management Protocol (SNMP) neither meet the performance requirements for dataplane tasks, nor provide a packet-consistent view of network state.

There have been numerous efforts to expose switch statistics through the dataplane, particularly to improve congestion management and network monitoring. One example is Explicit Congestion Notification in which a router stamps a bit in the IP header whenever the egress queue occupancy

exceeds a configurable threshold. Another example is IP Record Route, an IP option that enables routers to insert the interface IP address on the packet. Yet another example is Cisco’s Embedded Logic Analyzer Module (ELAM) [12] that traces the packet’s path inside the ASIC at layer 2 and layer 3 stages, and generates a summary to the network control plane. Instead of anticipating future requirements and designing specific solutions, we adopt a more generic, protocol-independent approach to accessing switch state.

10 Conclusion

We set out with a goal to rapidly introduce new dataplane functionality into the network. We showed how, by presenting a programmatic interface, using which end-hosts can query and manipulate network state directly using tiny packet programs. TPPs support both a distributed programming model in which every end-host participates in a task (e.g., RCP* congestion control), and a logically centralized model in which a central controller can monitor and program the network. We demonstrated that TPPs enable a whole new breed of useful applications at end-hosts: ones that can work *with* the network, have unprecedented visibility nearly instantly, with the ability to tie dataplane events to *actual* packets, unambiguously isolate performance issues, and act on network view without being limited by the control plane’s ability to provide such state in a timely manner.

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APPENDIX

Memory Map

In this section, we list the memory map to access switch statistics and per-packet metadata that is usually tracked by a switch. This list of statistics and metadata is by no means complete, nor necessarily tracked by all switches, but it serves as a reference list of statistics we found useful when implementing new dataplane tasks.

Counters and statistics

We borrow heavily from the list of counters from the OpenFlow 1.4 specification [32, Table 5]. We introduce more useful counters which would help monitoring and debugging the network. As switch vendors add more statistics, they could be made accessible to TPPs using a vendor-specific memory map. The vendor can then provide documentation on using the memory map using a data sheet.

When we refer to some values as ‘stats,’ or a ‘stats block’ we track the following four counters: packets, bytes, rate of packets, and rate of bytes. For example, the ‘Lookup stats’ block tracks the aggregate number of bytes and packets that resulted in a lookup on the table. This is useful to know for every table, as some packets (e.g., an ARP packet) may not match all tables (e.g., the layer 3 routing table).

Value Name	Purpose
Per ASIC. Namespace [Switch:]	
Switch ID	A unique ID given to a switch
Version Number	A global counter that tracks the generation of the switch forwarding state
Clock/uptime	The duration, in clock cycles, for which the switch has been online
Clock frequency	The clock frequency of the switching ASIC in cycles per second
Per Flow Table. Namespace [Stage <i>i</i> :] for the <i>i</i> th stage.	
Version Number	A per flow table version number that monotonically increases on every flow update
Reference Count	Number of active flow entries in the table
Lookup stats	Number of packets and bytes that resulted in a lookup in the table
Match stats	Number of packets and bytes that matched some entry in the table
Per Flow Entry. Namespace [FlowEntry <i>i</i> :] for the <i>i</i> th stage.	
Insert clock	The clock cycle when the flow entry was installed in the table
Match stats	Number of packets and bytes that resulted in a match on this entry
Per Port. Namespace [Link <i>i</i> :] for the <i>i</i> th link.	
Queued stats	Number of packets and bytes waiting on this port to be transmitted
Transmit stats	Stats block for packets transmitted on this port
Receive stats	Stats block for packets received on this port
Drop stats	Stats block for drops on this port
Error stats	Stats block for CRC/other bit errors on this port
Port status	Status bits (up/down/maintenance/etc.) for this port
Per Queue. Namespace [Queue <i>i</i> <i>j</i> :] for the <i>j</i> th queue on the <i>i</i> th link.	
Scheduling configuration block	This memory block contains counters and statistics that pertain to the scheduling algorithm (e.g., deficit round robin weight, quantum, etc.).
Queued stats	Stats block for packets currently queued
Transmit stats	Stats block for transmitted packets
Receive stats	Stats block for received packets
Drop stats	Stats block for dropped packets

Table 6: Statistics required from the ASIC, based on the standard values tracked by an OpenFlow 1.4 capable switch. These statistics must be accessible to the TPP.

Value Name	Purpose
Namespace: [PacketMetadata:].	
Input port	The input port number on which the packet arrived to the switch
Input port statistics	An indirection to the stats block tracking the port's counters
Output port bitmap	A bitmap indicating the port(s) out of which the packet will be forwarded. The bitmap is initialized to 0. If, at the end of the ingress pipeline, the bitmap is still set to 0, then the packet will be dropped.
Matched flow entry	An index into the flow entry that the packet matches at each table
Matched flow entry stats	Each flow entry tracks various statistics as shown above. This is an indirection to the stats block for the flow entry counters.
Enqueued queue ID	An index into the queue at the output port into which the packet is queued for transmission
Enqueued queue's stats	An indirection to the stats block tracking the queues's counters
Packet fields	All packet fields that were parsed by the parser for this packet. This includes all the standard OpenFlow 1.4 [32] fields, and our proposed fields for the TPP (§3.4).

Table 7: Per-packet metadata that must be available to every TPP at the ingress pipeline.

Value Name	Purpose
Namespace: [PacketMetadata:].	
Output port	The port through which the packet is currently being forwarded out of
Output port statistics	An indirection to the stats block tracking the port's counters
Output queue	The queue on which the packet was scheduled
Output queue statistics	An indirection to the stats block tracking the queue's counters

Table 8: Per-packet metadata that must be available to every TPP at the egress pipeline.

Per-packet metadata

Recall that the packet is processed once at the ingress pipeline, where its output port(s) is (are) determined. We expect the statistics listed in Table 7 to be available to every TPP. These statistics are not shared across packets. It is analogous to the `/proc/self` interface that a process can use to access its own statistics such as memory usage. Similarly, Table 8 lists statistics that should be available to every TPP at the egress.