### Advanced Topics in Communication Networks Programming Network Data Planes



Laurent Vanbever nsg.ee.ethz.ch

ETH Zürich Oct 29 2019



### Last week on

### Advanced Topics in Communication Networks

P4 hardware target

P4-based applications

How do we build a *fast* reprogrammable switch?



How can we allow network programmability in the field, at reasonable cost, and without sacrificing speed

supporting Tbps of backplane throughput

### Let's look at a concrete design: Reconfigurable Match Tables (RMT)



trol plane and the forwarding plane based on the approach known as "Match-Action". Roughly, a subset of packet bytes

#### **Categories and Subject Descriptors**

### [SIGCOMM'13]

### The paper argues that flexibility does not come at the price of performance or cost

### Outline

- Conventional switch chips are inflexible
- SDN demands flexibility...sounds expensive...
- How do we do it: The RMT switch model
- Flexibility costs less than 15%

### Enter...

### Reconfigurable Match Tables (RMT)

### Outline

- Conventional switch chip are inflexible
- SDN demands flexibility...sounds expensive...
- How do we do it: The RMT switch model
- Flexibility costs less than 15%

What kind of switch architecture could support flexibility and yet run at Terabits per second?

Throughput 1 Tbps aggregate Packet size 1000 bits average 10 *#* operations per packet (avg.) 10 billion op./second Requirements

Pipelined architectures organize processing through a sequence of processing units and local memory



### For flexibility, each processing unit/memory can be made generic



Each CPU can process distinct packets, with up to 10 packets going through the pipeline simultaneously



The runtime behavior of the parser & the match stages is defined through the RMT abstract model



How do we implement in hardware a programmable parser and a logical pipeline?

### How do we implement in hardware a programmable parser and a logical pipeline?



packets. A 10 Gb/s Ethernet link can deliver a new packet every 70 ns; a state-of-the-art Ethernet switch

ASIC with  $64 \times 40 \,\text{Gb/s}$  ports must process a new

nacket every 270 ps

[ANCS'13]

Parsing; Design principles; Reconfigurable parsers

#### 1. INTRODUCTION

Despite their variety *every* network device examines fields

## Parsing is the (complex) process of identifying and extracting the appropriate fields in a packet header

Throughput	Parser must run at line-rate parse 1 packet every 70 ns on a 10 Gbps link
Dependency	Parsing involves sequential processing as headers typically point to the next one
Incompleteness	Some headers do not even identify the subsequent header
Heterogeneity	Many header formats exist that can appear in various orders/locations

A parser can be divided into two separate blocks: header identification and field extraction



Source: Design Principles for Packet Parsers, Gibb et al.

In a programmable parser, the two modules rely on runtime information instead of hard-coded logic

stored in memory, e.g. in RAM and/or TCAM



Source: Design Principles for Packet Parsers, Gibb et al.

How do we implement in hardware a programmable parser and a logical pipeline? A compiler translates a given RMT logical pipeline (specified in P4) into a physical one



# The compiler maps each individual logical stage to one or more physical stage.



### The RMT pipeline in a few statistics

### **Our Switch Design**

- 64 x 10Gb ports
  - 960M packets/second
  - 1GHz pipeline
- Programmable parser
- 32 Match/action stages

- Huge TCAM: 10x current chips
  - 64K TCAM words x 640b
- SRAM hash tables for exact matches
  - 128K words x 640b
- 224 action processors per stage
- All OpenFlow statistics counters

## Building a RMT pipeline is only 15% more expensive than building a fixed-function switching pipeline

### Outline

- Conventional switch chip are inflexible
- SDN demands flexibility...sounds expensive...
- How do I do it: The RMT switch model
- Flexibility costs less than 15%

### The biggest cost is the memory... *not* the processing logic

### Cost of Configurability: Comparison with Conventional Switch

- Many functions identical: I/O, data buffer, queueing...
- Make extra functions optional: statistics
- Memory dominates area
  - Compare memory area/bit and bit count
- RMT must use memory bits efficiently to compete on cost
- Techniques for flexibility
  - Match stage unit RAM configurability
  - Ingress/egress resource sharing
  - Table predication allows multiple tables per stage
  - Match memory overhead reduction
  - Match memory multi-word packing

### That was just an academic paper Let's look at a real flexible pipeline



A small subset of our lab @ITET with two Tofino 3.2 Tbps, 32x 100 GbE QSFP28

That was just an academic paper Let's look at a real flexible pipeline



### Barefoot Tofino 6.5 Tbps backplane

several billion packets per second at line rate



### Barefoot Tofino 6.5 Tbps backplane

several billion packets per second at line rate



Tofino relies on Packet Header Vector (PHV) to pass states between stages



Tofino uses a folded pipeline in which the *same* stages are used for both the ingress and the egress pipeline



### What's next? Tofino 2: 12.8 Tbps (7 nm switching ASIC)



https://www.barefootnetworks.com/press-releases/barefoot-networks-unveils-tofino-2-the-next-generation-of-the-worlds-first-fully-p4-programmable-network-switch-asics/

### This week on

### Advanced Topics in Communication Networks

P4 hardware target P4-based applications

What cool things can we do with it? Data plane for programmability

Performance Monitoring Applications offloading

Platforms Correctness Management for Data plane programmability

	Language-Directed Network Perform	Hardware Denance Monite	esign for oring	,
		🚡 senata pdf (pag	e 1 of 15)	🗶 🔹 🔊 🛞 Q Search
ABS				
vetw				
wite				
ave				
ne c	Sonata: Quary D	ivon Stroo	ming N	atwork Talamatry
ives ic as	Soliata. Quel y-Di	iven strea	ning N	etwork referietry
xpre	Arnit Gunta	Rob Harr	ison	Marco Canini
arie	Princeton University	Princeton Uni	iversity	KAUST
W				
s ba	Nick Feamster	Jennifer Re	extord	Walter Willinger
wite	Princeton University	Princeton Un	iversity	NIKSUN Inc.
t lin nd s	ABSTRACT	:	1 INTRODU	JCTION
urge	Managing and securing networks requi	res collecting and	Network operato	rs routinely perform continuous monitor-
peer pat r	try systems do not allow operators to ex	e. Existing teleme-	o attacks. This n	nonitoring requires continuous, real-time
nly	queries needed to perform management or	scale to large traf-	neasurement an	d analysis-a process commonly referred
-	fic volumes and rates. We present Sonata,	an expressive and t	to as network tele-	metry [55]. Existing telemetry systems can
Na	and analysis of network traffic. Sonata pro	vides a declarative	ther support a	limited set of telemetry tasks [34, 40], or
	interface to express queries for a wide	range of common t	hey incur substa	ntial processing and storage costs as traffic
KE'	telemetry tasks; to enable real-time execu- tions each query across the stream proce	ition, Sonata parti- isor and the data	ates and queries	increase [7, 10, 58]. etry, pertame typically trade off scalabil-
śctw	plane, running as much of the query as it of	an on the network i	ty for expressiv	eness, or vice versa. Telemetry systems
см	switch, at line rate. To optimize the use	of limited switch t	hat rely on strea	m processors alone are expressive but not
rinis	that available resources focus only on traff	ic that satisfies the	calable. For exa DpenSOC [40] ca	mpie, systems such as NetQRE [58] and an support a wide range of queries using
cina.	query. Our evaluation shows that Sonata	can support a wide s	tream processor	s running on general-purpose CPUs, but
fonit 1-25	range of telemetry tasks while reducing th stream processor by as much as sman or	e workload for the t	hey incur substar	ntial bandwidth and processing costs to do
ups:	compared to existing telemetry systems.	acts of magnitude 2	nillion operation:	s per second for rates of 1 Tbps and packet
	CCS CONCEPTS	1	izes of 1 KB. Scal	ing to these rates using modern stream pro-
	Networks → Network monitoring		ressors is prohibit nagnitude) proce	ssing capacity per core [37, 39, 41, 59]. On
lacer	VEV4/ODDe	1	he other hand, tel	lemetry systems that rely on programmable
a the	KE I WORDS		witches alone ca	in scale to high traffic rates, but they give is to achieve this scalability. For example
publi pd/or	anaiyucs, programmable switches, stream	n processing	Marple [34] and	OpenSketch [56], can perform telemetry
JOCE	ACM Reference Format: Arnit Gunta Rob Harrison Marco Canini Nic	k Feamster Jennifer	asks by executir	ng queries solely in the data plane at line
e Co CM	Rexford, and Walter Willinger. 2018. Sonata: Q	uery-Driven Stream-	are, but the queri apabilities and n	tes that mey can support are limited by the nemory in the data plane.
Iped	ing Network Telemetry. In SIGCOMM '18: A Conference, August 20–25, 2018, Bodanest Huma	CM SIGCOMM 2018 arv. ACM, New York.	Rather than ac	cepting this apparent tradeoff between ex-
	NY, USA, 15 pages. https://doi.org/10.1145/32	30543.3230555	pressiveness and	scalability, we observe that stream proces-
			nodel; they both	apply an ordered set of transformations
	Permission to make digital or hard copies of all or personal or classroom use is granted without fee prov	r part of this work for ided that copies are not	over structured d	lata in a pipeline. This commonality sug-
	made or distributed for profit or commercial advanta	ge and that copies bear	sests that an oppo	ortunity exists to combine the strengths of
	of this work owned by others than ACM must be her	tored. Abstracting with	expressive querie	s, while still operating at line rate for high
	credit is permitted. To copy otherwise, or republish,	to post on servers or to	raffic volumes ar	id rates.
	redistribute to lists, requires prior specific permission	n and or a tee, Kennest		
	redistribute to lists, requires prior specific permissio permissions from permissions@acm.org.	n and/or a tee, kequest	To explore this work Traffic Area	s idea, we develop Sonata (Streaming Net-
	redistribute to lists, requires prior specific permission permissions from permissions@acm.org. SIGCOMM '18, August 20-25, 2018, Budapest, Hanga © 2018 Association for Computing Machiners:	n and/or a tee, kequest	To explore this work Traffic Ana system. Figure 1	s idea, we develop Sonata (Streaming Net- alysis), a query-driven network telemetry shows the design of Sonata: it provides

<ul> <li>A</li> </ul>	_≤ conext16.pdf (page 1 of 15)	🗶 = 🗇 🐵 Q, Search
LassDa	daw Fact Detection of Last D	aakata in Data
LOSSRA	dar: Fast Detection of Lost Pa	ackets in Data
	Center Networks	ans.off (sase 2 of 15)
		🗶 💌 👌 🛞 Q Search
ABSTR		
Packet loss		
caused by a		
network of		
detection o		
We also nee	FlowRadar: A Better N	etFlow for Data Centers
these losse		
are generic	Vuliang Li* Pui Mino* C	hanahaan Kim <sup>†</sup> Minlan Va*
fall short ii	*University of Southern Colif	anghout Kim Minian Tu
we propose	University of Southern Calif	onita Datejoor Networks
detection. 1		
individual I		
on prototyp		
is easy to i	Abstract	plement NetFlow in hardware is how to maintain an ac-
memory an	NetFlow has been a widely used monitoring tool with	time and space complexity. We need to handle collisions
analysis to:	a variety of applications. NetFlow maintains an active	during flow insertion and remove old flows to make room
with a few (	working set of flows in a hash table that supports flow insertion collision resolution and flow removing. This	for new ones. These tasks are challenging given the lim-
	is hard to implement in merchant silicon at data cen-	ited per-packet processing time at merchant silicon.
1. INT	ter switches, which has limited per-packet processing	no nancie unis chaitenge, today s Netriow is imple- mented in two ways: (1) Using complex custom silicon
Packet loss	time. Therefore, many NetFlow implementations and	that is only available at high-end routers, which is too
happen for	subset of nackets to monitor. In this namer, we observe	expensive for data centers; (2) Using software to count
for one yes	the need to monitor all the flows without sampling in	sampled packets from hardware, which takes too much
losses, 4 n	short time scales. Thus, we design FlowRadar, a new	able NetFlow in data centers, operators have to mirror
random cor	way to maintain flows and their counters that scales to a large number of flows with small memory and bandwidth	packets based on sampling or matching rules and ana-
immediate	overhead. The key idea of FlowRadar is to encode per-	lyze these packets in a remote collector [26, 40, 44, 34].
especially y	flow counters with a small memory and constant inser-	It is impossible to mirror all the packets because it takes
	tion time at switches, and then to leverage the computing	storage and computing resources at the remote collector
Permission to a	power at the remote collector to perform network-wide decoding and analysis of the flow counters. Our analysis	to analyze every packet. (Section 2)
distributed for	uation shows that the memory usage of FlowRadar is	However, in data centers, there is an increasing need
and the full cit owned by othe	close to traditional NetFlow with perfect hashing. With	to have visibility of the counters for all the flows all the
mitted. To cop lists, requires #	FlowRadar, operators can get better views into their net-	sient loops, blackholes, and switch faults that only hap-
permissions@2	works as demonstrated by two new monitoring applica- tions we build on ton of FlowPadar.	pen to a few flows in the Network and to perform fine-
CONEXT T	tions we build on top of a sourcease.	grained traffic analysis (e.g., anomaly detection). We
DOE: http://dp	1 Introduction	need to cover these nows all the time to identify transient lower bursts and attacks in a timely fashion (Section 3)
	1 Introduction	In this paper, we propose FlowRadar, which keeps
	NetFlow [4] is a widely used monitoring tool for over 20	counters for all the flows with low memory overhead
	years, which records the flows (e.g., source IP, destina-	and exports the flow counters in short time scales (e.g.,
	their properties (e.g., packet counters, and the flow start.	to ms). The key design of PlowRadar is to identify the best division of labor between chean switches with lim-
	ing and finish times). When a flow finishes after the in-	ited per-packet processing time and the remote collector
	active timeout, NetFlow exports the corresponding flow	with plenty of computing resources. We introduce en-
	records to a remote collector. NetFlow has been used for	coded flowsets that only require simple constant-time in-
	a variety or monitoring applications such as accounting network usage, canacity planning, troubleshooting, and	structions for each packet and thus are easy to implement with merchant silicon at chean switches. We then decode
	attack detection.	these flowsets and perform network-wide analysis across
	Despite its wide applications, the key problem to im-	time and switches all at the remote collector. We make

INT-current-spec.pdf (page 1 of 28) ▲ INT-current-spec.pdf (page 1 of 28) ▲ ● ● ● ● ● ● ● ● ● ● ● ● ● ● ● ● ● ● ●
In-band Network Telemetry (INT)
Changhoon Kim, Parag Bhide, Ed Doe: <i>Barefoot Networks</i> Hugh Holbrook: <i>Arista</i> Anoop Ghanwani: <i>Dell</i> Dan Daly: <i>Intel</i> Mukesh Hira, Bruce Davie: <i>VMware</i>
Introduction Terms What To Monitor Switch-level Information
Ingress Information Egress Information Buffer Information Processing INT Headers INT Header Types
Handling INT Packets Header Format and Location INT over any encapsulation On-the-fly Header Creation Header Format Header Location and Format INT over Geneve



	Network Perform	nance Mon	itoring		
			(page 1 of 16)		
AB2 Netw					
	Sonata: Ouerv-Dr	iven Stre	aming N	etwork Telemetry	
	Arpit Gupta	Rob Ha	arrison	Marco Canini	
warie				KAUST	
	Nick Feamster	Jennifer	Rexford	Walter Willinger	
		Princeton		NIKSUN Inc.	
	ABSTRACT		1 INTRODU	JCTION	
	Managing and securing networks requir	es collecting and	Network operato	rs routinely perform continuous monitor-	
	queries needed to perform management or	scale to large traf-	measurement an	d analysis-a process commonly referred	
Net	and analysis of network traffic. Sonata prov	rides a declarative		limited set of telemetry tasks [34, 40], or	
KE'	tions each query across the stream proce			etry systems typically trade off scalabil-	
ACM Srinh	memory, Sonata dynamically refines each		scalable. For exa	mple, systems such as NetQRE [58] and	
	range of telemetry tasks while reducing the	workload for the	they incur substa	ntial bandwidth and processing costs to do	
	CCS CONCERTS			ing to these rates using modern stream pro-	
	Networks → Network monitoring				
	VPVWODDe		the other hand, te	lemetry systems that rely on programmable	
	RETWORDS analytics programmable switcher stream				
				OpenSketch [56], can perform telemetry	
	ACM Reference Format: Arpit Gupta, Rob Harrison, Marco Canini, Niel				
	Rexford, and Walter Willinger. 2018. Sonata: Quing Network Telemetry. In SECOMM '18: 40			nemory in the data plane.	
			sors and program	mable switches share a common processing	
	Permission to make digital or hard copies of all or			apply an ordered set of transformations late in a nineline. This commonality sur-	
				in a single telemetry system that supports s, while still constating at line rate for high	
				id rates. s idea, we develop Sonata (Streaming Net-	
			traffic volumes ar To explore this work Traffic Ana system. Figure 1	a rates. s idea, we develop Sonata (Streaming Net- ilysis), a query-driven network telemetry shows the design of Sonata: it provides	

LossRad	lar: Fast Detection of Lost P Center Networks	ackets in Data
ABSTR. Packet loss canned by <i>i</i> and have <i>vi</i> , network, op detection o We also nec information these losse are generic fall short il low overhe we propose detection. A individual 1	FlowRadar: A Better N Yuliang Li* Rai Miao* C *University of Southern Caliy	etFlow for Data Centers hanghoon Kim <sup>1</sup> Minlan Yu <sup>*</sup> ornia "Barefoot Networks
	Abstract	plement NetFlow in hardware is how to maintain an ac-
	NetFlow has been a widely used monitoring tool with	time and space complexity. We need to handle collisions
analysis to:		during flow insertion and remove old flows to make room
with a few c	insertion, collision resolution, and flow removing. This is hard to implement in merchant silicon at data cen- ter switches, which has limited per-packet processing	for new ones. These tasks are challenging given the lim- ited per-packet processing time at merchant silicon. To handle this challenge, today's NetFlow is imple- mented in two wave: (1) Union complex content vilicon
Packet loss	time. Therefore, many NetFlow implementations and other monitoring colutions have to complete a called a	that is only available at high-end routers, which is too
		expensive for data centers; (2) Using software to count
	the need to monitor all the flows without sampling in	
losses, 4 n		able NetFlow in data centers, operators have to mirror
		packets based on sampling or matching rules and ana-
	overhead. The key idea of FlowRadar is to encode per-	
	flow counters with a small memory and constant inser- tion time at switches, and then to leverage the comparing	too much bandwidth to mirror the traffic, and too many
		storage and computing resources at the remote collector
	decoding and analysis of the flow counters. Our eval-	
	uation shows that the memory usage of FlowRadar is	
		time. We need to cover all the flows to capture those tran-
		sient loops, blackholes, and switch faults that only hap-
	tions we build on top of FlowRadar.	pen to a few flows in the Network and to perform fine-
	1 Introduction	losses, bursts, and attacks in a timely fashion. (Section 3) In this paper, we propose FlowRadar, which keeps
		counters for all the flows with low memory overhead
	ing and finish times). When a flow finishes after the in-	ited per-packet processing time and the remote collector
	active timeout, NetFlow exports the corresponding flow	with plenty of computing resources. We introduce en-
		coded flowsets that only require simple constant-time in-
	attack detection.	these flowsets and perform network-wide analysis across

Search
tworks
i Dell
r: Intel
Mware

Dappe	er: Data Plane P	Performance Diagn	osis of TCP		
moj					
ABSTRAC					
With more app used to diagno	Netw	ork-Wide Heavy	Hitter Detec	tion with	
		Commodi	ty Switches		
	Ro	b Harrison, Qizhe Cai, Arp Princetor	pit Gupta, and Jennife University	r Rexford	
ection is lim eting for shar or the receiver	ABSTRACT Many network monito	ring tasks identify subsets of traffic	60 -	5m Interval	
	Protocol Independent S identify these "heavy h		(§) 40 -	10s Interval	
		statistics across packets and compar- 1. However, network operators often sting traffic on a <i>network-wide</i> basis.	<sup>2</sup> 20 -		
icavier-weight		een line-rate monitoring and network- sent a distributed heavy-hitter detec- ks modeled as one-big switch. We use		20 30 40 50 ampling Rate (1/x)	
Performance D			Figure 1: This graph shows between two major ISPs [12] 4		
vork performation protocols	traces. We demonstrate tect network-wide hea communication overhead	e that our solution can accurately de- vy hitters with up to 70% savings in ad compared to an existing approach	under high sampling rates, re- the monitoring interval decree		
I. INTRO	with a provable upper	bound.	solutions use approximate ory and processing overh accuracy, in order to deal w		em- i in ble
	Network operators offe traffic, to detect attack	n need to identify outliers in network s or diagnose performance problems.	on the switches. While prior work has f		ion
		ct unusual traffic is to perform "heavy dentifies the top-k flows (or flows ex- ed threshold), according to some met-	an a single switch, network network-wide heavy hitter and superspreaders [27] o		
age and that co uge. Copyrigh han ACM must			monitored only at one loc separately at each switch not sufficient. Large flows		ers s is lar"
	or TCP incast [4] in re- heavy-hitter detection				
		ogrammable switches open up new ating traffic statistics and identifying the data plane [17, 18, 24, 27]. These	communication overhead dinator.		
			often resort to sampling i defined based on the needs	/x packets, where x is operat of the specific network. Howe betweet all a set	
					. In nile
			time intervals and it quic decrease. In modern data		
SketchLe Measur	earn: Relievin ement with A	g User Burdens i utomated Statist	n Approxima ical Inference	te	
SketchLe Measur State Ke ABSTRAC Setwork me	earn: Relievin <sub>i</sub> ement with A	g User Burdens in utomated Statist	n Approxima ical Inference	te 2 - 2 8 9 9 1000	
SketchLe Measur *state k ABSTRAC Network me yource requir yource avain are the right high up we he neasurement	earn: Relievin ement with A	g User Burdens i utomated Statist - entropy Sketch: Adaptive Measur	n Approxima ical Inference resource and Fast Net ements	te 2 - 0 - 0	
SketchLe Measur 'State Ke ABSTRAK Network me source roui While approp source roui while approp uonee roui while approp isch user be neasuremen liekts when resources. In arations wil	earn: Relieving ement with A costs Elastic :	g User Burdens i utomated Statist setters sketch: Adaptive Measur atci o	n Approximai ical Inference superior and Fast Net ements	te Z • D D D two work-wide	
SketchLe Measur 'Sate K 'State K ABSTRA Kärwark nei wähle approximitette Kathuren konserentere kathuren konserentere kathuren konserentere kathuren konserentere kathuren konserentere kathuren konserentere kathuren konserentere kathuren konserentere kathuren konserentere	Elastic :	g User Burdens i utomated Statist setter: Sketch: Adaptive Measur	n Approximat ical Inference "montres" and Fast Net ements	te Z · O O Como work-wide	
SketchLee Measure 'State Ke 'State Ke ABSTRIA Kowak ng Wala ayan base awin are the right State sket Bets when reading and the state facts water facts here is facts here i	Elastic : Provide the second	g User Burdens in utomated Statist contract Sketch: Adaptive Measur at s to	n Approximat ical Inference newstarts and Fast Net ements	te 2 0 0 0 000 work-wide	2 - 2
SketchLec Measure 'state ki 'state ki Mail agent biological management har non-service management har non-service management har non-service har non-service h	Elastic S	g User Burdens in utomated Statist settores Sketch: Adaptive Measur acts o C Rethinking I	n Approximai ical Inference and Fast Net ements execution	te 2 - (c) = 0 ++++ work-wide mart (her) + (1) Rule Them A Monitoring v	۲۰۵ II: vith U
SketchLe Measur *bate *bate More to More to Mo	Elastic S Barni A	g User Burdens in utomated Statist setting Sketch: Adaptive Measur a.c. (a) O Rethinking M	and Fast Net ements	te vork-wide work-wide merener (*) Rule Them A Monitoring v	(Z → ∅) Ill: Vith U
SketchLe Measure 'state K 'state K ABSTRM Network neg state state Sath und the Sath	Elastic S Elastic S File yungten Mintin Tic hunn Mintin Mi	g User Burdens i utomated Statist setters Sketch: Adaptive Measur aus o Rethinking f	an Approximatical Inference and Fast Net ements • spentaria the Sketch to Vetwork Flow	te work-wide mortany 1/10 Rule Them A Monitoring v orsangeri , Vyas Sekar - Carnege Malon Univ	Z • 2 Ill: vith U . Vladimir
SketchLc Measure 'state St 'state St State St State St State St State St State St State St State St State St St State St St St St St St St St St St St St St S	Elastic S Flastic S Elastic S Flastic S Flasti	g User Burdens i utomated Statist setters Sketch: Adaptive Measur statist Zaoxing Liur, Antoni Zaoxing Liur, Antoni Sastract	and Fast Net ements une Sketch to Verwork Flow	te work-wide Rule Them A Monitoring v orsanger', Vyas Sokar Camego Mason Univ	Z = 0 III: vith U ersity
SketchLec Measure 'state & 'state & ATSTRICK Karwaka and	Elastic S Flastic S Elastic S Very suggest Control Very suggest Control	g User Burdens i utomated Statist setting Sketch: Adaptive Measur Casta o Rethinking P Zaoxing Lui, Antoni data and ABSTRACT	an Approximatical Inferences amore and Fast Net ements reported to the second s	te work-wide work-wide Rule Them A Monitoring v orsanger', Vyas Sokar "Carnege Malon Univ	III: vith U vith u multi-facetote
SketchLeck Measure "state it "state it "state it "state it "state it state	Elastic S Flastic S	g User Burdens i utomated Statist setting Sketch: Adaptive Measur CLS G C Rethinking M Zaxing Lir, Antori 1.0 ABSTRACT	and Fast Net ements and Fast Net ements • second and Network Flow Manousis-, Grogory V Manousis-, Grogory V	te work-wide work-wide Rule Them A Monitoring v ortanger, Vyas Sokar Carrege Maken Univ	III: With U Multi-faceted multi-f
SketchLeck Measure "state it "state it "state it "state it "state it sketter beinger sketter b	Elastic S Filestic S Filesti	g User Burdens i utomated Statist setting Sketch: Adaptive Measur Cost of CRethinking M Zaoxing Lir, Antoni Cast of C ABSTRACT	and Approximati ical Inference and Fast Net ements • • • • • • • • • • • • • • • • • • •	te work-wide work-wide Rule Them A Monitoring v oranger, Vyss Soker Carrege Melon dww I Introduction News Canager, Vyss Soker Carrege Melon dww I Introduction News Canager, Vyss Soker Carrege Melon dww Melon	Z = 2 III: vith U undining find expression prices accurate have himse
SketchLeck Measure **stare it **stare it **s	Elastic S references Elastic S references referenc	g User Burdens i utomated Statist setting Sketch: Adaptive Measur acts to CRethinking I Zaoxing Lur, Antoni Casta Casta ABSTRACT	an Approximati ical Inference file of the and Fast Net ements • spent as • sp	te work-wide work-wide Rule Them A Monitoring w oranger, Vyss Seiar Carregio Mellon Univ I Introduction Rundenter Statistics Seisen affere and seisen and seisen affere term of task including to provide the seisen affere term of task including to provide the seisen affere term of task including to provide the seisen affere term of task including to term	Z = 0 III: Vith U Vidadinai pers account and forema- pers account and forema- pers account and forema- tains engine the second of the second of
SketchLeck Measure ************************************	Elastic : Figure and the second seco	g User Burdens i utomated Statist setting Sketch: Adaptive Measur Construction Categories Categorie	and Fast Net ements and Fast Net ements · • ***********************************	te work-wide work-wide work-wide transfer	Il:     Idi     I
Statestate a state at a state at at a state at at a state at at at at at at at at at at	Elastic : Figure 1 Elastic : Figure 1 Figure 1 Figu	g User Burdens i utomated Statist servers Sketch: Adaptive Measur at a c	an Approximati ical Inference and Fast Net ements . • • • • • • • • • • • • • • • • • • •	te work-wide work-wide wardsart to Rule Them A Monitoring v organger, Vyas Sokar Carnegie Melion Univ I Induction Data and the formation of the formation of the formation of the formation of the formation of the formation of the formation of the formation of the formation of the formation of the formation o	II: With U Viadimir ensity with a differentiation and formation and formation and formation throw changes to be a provided by the provided by the set of the set of the set of the set of the set of the set of the
Sherebener The second	Elastic : Filestic : Filesti	g User Burdens i utomated Statist :::::::::::::::::::::::::::::::::::	an Approximation ical Inferences and Fast Net ements 	te more and the second of the	X = 0 
Shertcherd Measure **ara to **ara to ***ara to **ara	Elastici Transitional Sectors of the sectors of the sectors of the sectors of the sector of the sector of the sectors of the s	g User Burdens i utomated Statist eternet Sketch: Adaptive Measur To Categories Categori	an Approximation in the second	te more to the special of the speci	If is a set of the set of th
Sherebelle Measure <sup>1</sup> shara ta <sup>1</sup> shara t	Elastic : Filestic : Filesti	g User Burdens i utomated Statist : :::::::::::::::::::::::::::::::::	an Approximati ical Inference rements and Fast Net ements reme	te Control of the second seco	If the second seco
Sherebener The second	earn: Relieving ement with A	g User Burdens i utomated Statist control of the second statistical of the second statistical of the second statistical of the second control of the secon	an Approximati ical Inference rements and Fast Net ements rements rements mension research and rement second and rements remen	te Contraction of the second	III: Vitable University Vitable University Naddimine Segment Segm
Statistical and a statistical		g User Burdens i utomated Statist control of the second state of t	an Approximati ical Inference rements and Fast Net ements rements methods and the second of the second the second second rements of the second the second second rements of the second the second second second rements of the second rements of the second the second second second rements of the second rements o	te Contract of the second sec	III: Vitadiment vitatione vita
Statistical and a statistical		g User Burdens i utomated Statist setting Sketch: Adaptive Measur 3133 13 Control Rethinking f Zaodg Lu, Atog Abbreve	an Approximati ical Inference and Fast Net ements • ements • emerse ments • emerse •	te Control to the control to the co	Z • 12 III: UIII: VIAdimini Vitadimini Anayana Anayan Anayana Anayana Anayana Anayana Anayana Anayan
Statistical and a statistical		g User Burdens i utomated Statist steams steams steams and a company steams and a company ste	an Approximation in a second s	te Contrevente C	II: UNIT OF A CONTRACT OF A CO
Skatchele States Sta		g User Burdens i utomated Statist Control of the second Sketch: Adaptive Measur Control of the second Control	an Approximation in a second s	te Contrevente C	II: UNIT OF THE STATE OF THE ST
### Current monitoring methods are inadequate

- Not fast enough
  - Involve CPU and control planes
  - Network state changes rapidly

### Do not provide end-to-end state

 Difficult to correlate per-element state with the actual path of a flow

### INT : In-band Network Telemetry

- Mechanism for collecting network state in the dataplane
  - As close to realtime as possible
  - At current and future line rates
  - With a framework that can adapt over time
- Examples of network state
  - Switch ID, Ingress Port ID, Egress Port ID
  - Egress Link Utilization
  - Hop Latency
  - Egress Queue Occupancy
  - Egress Queue Congestion Status
  - .....

			4 B	ytes		
Ve	r	Flags	Instruction Count	Max Hop Count	Total Hop Count	Metadata
Ins	struc	tion Bitma	ip	Reserved		Header
0			Most Recent I	NT Metadata		
0			INT Me	tadata		
						Metadata
1			First INT N	Vetadata		

# INT using P4

- P4 enables flexible packet parsing and modification for INT
- P4 allows INT to adapt to
  - Any Encapsulation format
  - Any State required to be collected
  - Any feature, protocol current and future



### HULA: INT + Flowlet routing

### 1. Periodic INT probes

disseminate path utilization to switches

# 2. Flowlet detection and path selection

- happens at all switches
- hop-by-hop adaptive routing









- INT provides real-time network state directly in the dataplane
  - Scales to arbitrarily large networks
  - Scales to current and future link speeds
  - Can adapt to any network, any encap, any application
- Knowledge of real-time network state opens up new possibilities
  - Enhanced monitoring and troubleshooting
  - Network-state aware routing
  - •

ABS		scrata odf (page 1 of 15)	
ABS			🗶 💌 🗗 🛞 Q, Starth
Netw			
endp			
have			
he c	Sonata: Query-Dri	ven Streaming N	letwork Telemetry
ve at	Soliata Quely Di	ren streaming i	termonic referinction y
expre	Arpit Gupta	Rob Harrison	Marco Canini
varie	Princeton University	Princeton University	KAUST
milia	Nick Feamete-	Jannifar Payford	Walter Willinger
s ba	Princeton University	Princeton University	NIKSUN Inc
wite at lin	ADOTD A OT		Increase and a second s
and s	ABSTRACT Menoring and comming activity for	1 INTROL a collecting and Natural areas	JUCTION
arge	analyzing network traffic data in real time.	Existing teleme- ing to track eve	nts ranging from performance impairments
hat «	try systems do not allow operators to expr	ess the range of to attacks. This	monitoring requires continuous, real-time
only	queries needed to perform management or s fic volumes and rates. We present Sonata a	cale to large traf- measurement a to as network to	and analysis—a process commonly referred
CC	scalable telemetry system that coordinates	joint collection collect and anal	lyze measurement data in real time, but they
Net	and analysis of network traffic. Sonata provi	des a declarative either support	a limited set of telemetry tasks [34, 40], or
	telemetry tasks; to enable real-time executi	on, Sonata parti- rates and queri	es increase [7, 10, 58].
KE	tions each query across the stream proces	sor and the data Existing tele	metry systems typically trade off scalabil-
	plane, running as much of the query as it can rewitch at line rate. To optimize the use of	ton the network ity for express f limited switch that rely on str	aveness, or vice versa. Telemetry systems
ACM .	memory, Sonata dynamically refines each	query to ensure scalable. For e	xample, systems such as NetQRE [58] and
ienka	that available resources focus only on traffic	that satisfies the OpenSOC [40]	can support a wide range of queries using
Kim. Monit	ranze of telemetry tasks while reducing the	workload for the they incur subs	tantial bandwidth and processing costs to do
21-2:	stream processor by as much as seven orde	rs of magnitude so. Large netwo	orks can require performing as many as 100
and as a	compared to existing telemetry systems.	million operation sizes of 1 KB. Sc	ons per second for rates of 1 Tbps and packet aling to these rates using modern stream pro-
	CCS CONCEPTS	cessors is prohi	bitively costly due to the lower (2-3 orders of
Permi	<ul> <li>Networks → Network monitoring;</li> </ul>	magnitude) pro	cessing capacity per core [37, 39, 41, 59]. On
or pro	KEYWORDS	switches alone	can scale to high traffic rates, but they give
arthor (1997)	analytics, programmable switches, stream j	processing up expressiven	ess to achieve this scalability. For example,
ndice	ACM Reference Format:	Marple [34] an	d OpenSketch [56], can perform telemetry ting queries colely in the data plane at line
2011	Arpit Gupta, Rob Harrison, Marco Canini, Nick	Feamster, Jennifer rate, but the qu	eries that they can support are limited by the
KCM	Restord, and Walter Willinger. 2018. Sonata: Que ing Network Telemetry. In SIGCOMM '18: ACI	ry-Driven Stream- d SIGCOMM 2018 capabilities and	memory in the data plane.
utex	Conference, August 20–25, 2018, Budapest, Hungar	y. ACM, New York, Rather than :	accepting this apparent tradeoff between ex- d scalability, we observe that stream proces-
	NY, USA, 15 pages. https://doi.org/10.1145/3230	543.3230555 sors and progra	mmable switches share a common processing
	Permission to make digital or hard copies of all or u	art of this work for model; they bo	th apply an ordered set of transformations
	personal or classroom use is granted without fee provid	ed that copies are not red that copies are not rests that an or	a data in a pipeline. This commonality sug- poortunity exists to combine the strengths of
	move or distributed for profit or commercial advantage this notice and the full citation on the first page. Copyri	ghts for components both technologi	es in a single telemetry system that supports
	of this work owned by others than ACM must be honor credit is permitted. To come otherwise, or percellish to	ed. Abstracting with expressive quer	ies, while still operating at line rate for high
	redistribute to lists, requires prior specific permission a	ind/or a fee. Request To	and rates.
		10	his idea, we develop Sonata (Streaming Net-
	permissions from permissions@acm.org. SIGCOMM '18. Aurunt 20-25. 2018. Budowert Honsory	work Traffic A	nalysis), a query-driven network telemetry
	permissions from permissions@acm.org. SIGCOMM '18, August 20-25, 2018, Budapest, Hangary © 2018 Association for Computing Machinery.	work Traffic A system. Figure	is idea, we develop Sonata (Streaming Net- nalysis), a query-driven network telemetry 1 shows the design of Sonata: it provides staffage that an groups groups for constants

Q. (1)	C connected fields and st	🗶 💌 👌 🛞 Q, Search
LossRa	dar: Fast Detection of Lost P Center Networks	ackets in Data
E* Q		🗶 💌 👌 🛞 Q, Search
ABSTR.		
Packet loss		
and have si		
network or		
detection o		
We also nee	FlowRadar: A Better N	etFlow for Data Centers
these losse		
are generic	VE DINE DINE	the second se
fall short is	rullang Li Rul Milao C	nangnoon Kim Minian tu
low overne:	University of Southern Cauj	fornia Barefoot Networks
detection. V		
individual I		
entire netwo		
is easy to i	Abstract	plement NetFlow in hardware is how to maintain an ac-
memory an	NetFlow has been a widely used monitoring tool with	tive working set of flows using a data structure with low
information	a variety of applications. NetFlow maintains an active	time and space complexity. We need to handle collisions during flow insertion and remove old flows to make room
with a few s	working set of flows in a hash table that supports flow	for new ones. These tasks are challenging given the lim-
	insertion, collision resolution, and flow removing. This is hard to implement in marshant cilizon at data can	ited per-packet processing time at merchant silicon.
1 INT	ter switches, which has limited per-nacket processing	To handle this challenge, today's NetFlow is imple-
Desket land	time. Therefore, many NetFlow implementations and	mented in two ways: (1) Using complex custom silicon that is only available at high and postage, which is too
happen for	other monitoring solutions have to sample or select a	expensive for data centers; (2) Using software to count
keynote [7]	subset of packets to monitor. In this paper, we observe	sampled packets from hardware, which takes too much
for one yes	short time scales. Thus, we design FlowRadar, a new	CPU resources at switches. Because of the lack of us-
random cor	way to maintain flows and their counters that scales to a	able NetFlow in data centers, operators have to mirror packets based on sampling or matching rules and ana-
immediate	large number of flows with small memory and bandwidth	lyze these packets in a remote collector [26, 40, 44, 34].
significantl	overnead. The key idea of Flowkadar is to encode per- flow counters with a small memory and constant inser-	It is impossible to mirror all the packets because it takes
especially v	tion time at switches, and then to leverage the computing	too much bandwidth to mirror the traffic, and too many
Permission to a	power at the remote collector to perform network-wide	to analyze every packet. (Section 2)
or classroom a distributed for	decoding and analysis of the flow counters. Our eval-	However, in data centers, there is an increasing need
and the full cit.	uation snows that the memory usage of FlowKadar is close to traditional NetFlow with nerfect hashing. With	to have visibility of the counters for all the flows all the
owned by othe mitted. To cop	FlowRadar, operators can get better views into their net-	time. We need to cover all the flows to capture those tran-
lists, requires p permissions@a	works as demonstrated by two new monitoring applica-	stent toops, btackholes, and switch faults that only hap- nen to a few flows in the Network and to perform fine-
CoNEXT 'I	tions we build on top of FlowRadar.	erained traffic analysis (e.g., anomaly detection). We
© 2016 ACM.		need to cover these flows all the time to identify transient
LOC MIN/M	1 Introduction	losses, bursts, and attacks in a timely fashion. (Section 3)
	NatElaw [4] is a widah usad monitoring tool for over 20	In this paper, we propose FlowRadar, which keeps
	years, which records the flows (e.g., source IP. destina-	and exports the flow counters in short time scales (e.g.
	tion IP, source port, destination port, and protocol) and	10 ms). The key design of FlowRadar is to identify the
	their properties (e.g., packet counters, and the flow start-	best division of labor between cheap switches with lim-
	ing and finish times). When a flow finishes after the in-	ited per-packet processing time and the remote collector with electric of computing measures. We introduce or
	records to a remote collector. NetFlow has been used for	coded flowsets that only require simple constant-time in-
	a variety of monitoring applications such as accounting	structions for each packet and thus are easy to implement
	network usage, capacity planning, troubleshooting, and	with merchant silicon at cheap switches. We then decode
	attack detection.	these flowsets and perform network-wide analysis across
-	pespite its wide applications, the key problem to im-	ume anu switches all at the remote collector. We make

🔟 💌 👌 🔕 Q. Search
In-band Network Telemetry (INT)
June 2016
Changhoon Kim, Parag Bhide, Ed Doe: Barefoot Networks
Hugh Holbrook: Arista
Anoop Ghanwani: Dell
Dan Daly: Intel
Mukesh Hira, Bruce Davie: VMware
Introduction
Terms
What To Monitor
Switch-level Information
Ingress Information
Egress Information
Builer Information
INT Header Types
Handling INT Packets
Header Format and Location
INT over any encapsulation
On-the-fly Header Creation

Dappe	er: Data Plane F	Performance Diagn	osis of TCP		
mol	990	🚡 sosri8-final3i	6 pdf (page 1 of 7)	🗶 💌 🎒 🛞 Q, Search	
тој					
STRAC					
more app to diagne	Netw	ork-Wide Heavy	Hitter Detec	tion with	
ing the er rights to		Commodit	ty Switches		
ad, our sj ime <i>near</i> p-of-rack	Ro	b Harrison, Qizhe Cai, Arp Princeton	it Gupta, and Jennife <sup>University</sup>	er Rexford	
on is lim g for shar	ABSTRACT Many network monitor	ring tasks identify subsets of traffic	60 -	5m Interva	
es now o mmodity	that stand out, e.g., to Protocol Independent S identify these "heavy h	-k flows for a particular statistic. A witch Architecture (PISA) switch can itter" flows directly in the data plane,	<u>କ୍</u> ରି 40 -	50s Interva 10s Interva	
e Dapper c. To redu	by aggregating traffic ing against a threshold want to identify interes	statistics across packets and compar- However, network operators often	20 - C		
lightweig r-weight	To bridge the gap betwee wide visibility, we pre-	en line-rate monitoring and network- ent a distributed heavy-hitter detec-	0 2 10	20 30 40 50 ampling Rate (1/x)	-
words mance E	adaptive thresholds to ing directly in the data	ks modeled as one-big switch. We use perform efficient threshold monitor- plane. We implement our system us-	Figure 1: This graph show	the recall for detecting heavy	∼hitters
Concept perform:	ing the P4 language, ar traces. We demonstrate tect network-wide bes	d evaluate it using real-world packet that our solution can accurately de- vy hitters with up to 70% savings in	between two major ISPs [12] w under high sampling rates, re the monitoring interval decre	ith different monitoring intervi call quickly diminishes and we ases.	ils. Even rsens as
NTRC	communication overhe with a provable upper	ad compared to an existing approach bound.	solutions use approximate	data structures, that boun	i mem-
ic clouc on perfe	1 INTRODUCT Network operators ofte	ION n need to identify outliers in network	accuracy, in order to deal w on the switches.	with the limited resources av	ailable
ion to m	traffic, to detect attack A common way to dete hitter" detection that is	s or diagnose performance problems. et unusual traffic is to perform "heavy lentifies the ton-k flows (or flows ex-	While prior work has f at a single switch, network network-wide heavy hitter	ocused on heavy-hitter de k operators often need to tr s. For example, port scanne	ack the rs [15]
onal or a ne not n I that coj	ceeding a pre-determin ric. For example, netwo	ed threshold), according to some met- rk operators often track destinations	and superspreaders [27] c monitored only at one loc separately at each switch	ould go undetected if the ta ation. Detecting the heavy and then combining the re	affic is hitters
Opyrigh M must otherwi	receiving traffic from a high-precision in order or TCP incast [4] in re	large number of distinct sources with to detect and mitigate DDoS attacks al time. In traditional networks, this	not sufficient. Large flow at multiple locations but s	s can easily fall "under the till have sizable total volu	radar" ne. Ap-
s from P 7, April ACM_1	heavy-hitter detection or flow logs [5, 6]. Pr possibilities for any	relies on analyzing packet samples ogrammable switches open up new ating traffic statistics and identifici-	prying a lower detection the chance of missing lar communication overhead	unresnow at each switch r ge flows, at the expense of to report counts to a centr	eauces higher al coor-
ttp://dx.d	large flows directly in Permission to make digital	the data plane [17, 18, 24, 27]. These or hard copies of all or part of this work for	dinator. Additionally, networks often resort to sampling	that forward high traffic v	olumes erator-
	personal or classroom use is p made or distributed for profit this notice and the full citation	ranted without fee provided that copies are not or commercial advantage and that copies bear n on the first page. Copyrights for components	defined based on the needs sampling can result in su	s of the specific network. He abstantially reduced accur	owever, acy on
	of this work owned by other credit is permitted. To copy o redistribute to lists, requires permissions from permission	than ACM must be honored. Abstracting with therwise, or republish, to post on servers or to prior specific permission and/or a fee. Request softwarm cert	Figure 1, we show the impo performing heavy-hitter	detection on a link betwe	y while en two
	SOSR '18, March 28–29, 2018, © 2018 Association for Com ACM KBN 978-1-2503-5664	Los Angeles, CA, USA suting Machinery. 0.1809. \$15.00	major ISPs [12] processin Even with high sampling time intervals and it quic	g approximately 1 GBps of rates, recall is quite low o kly diminishes as samplin	traffic. 1 short g rates
	https://doi.org/10.1145/3185	467.3185476	decrease. In modern data	center networks where sw	vitches
etchLe 1easur	earn: Relievin rement with A	g User Burdens in utomated Statist	Approximatical Inference	te 2	
etchLe Aeasur	earn: Relievin rement with A	g User Burdens in utomated Statisti	7 - 0 0 0 www. n Approxima ical Inference	te s	
etchLe 1easur ate Ke IRA(	earn: Relievin rement with A	g User Burdens i tutomated Statisti	7 + 0 0 0 www	te 2 → 2 © L sut	
etchLe leasur te Ke RA( Reime requi avrian	earn: Relievin ement with A	g User Burdens in utomated Statisti - attracts	Approxima ical Inference report 10	te 2 work-wida	
etchLe easur RA( k me avin avin avin ight er bu	earn: Relievin rement with A	g User Burdens in utomated Statisti • etternet Sketch: Adaptive Measur	Approxima ical Inference report M	te Service work-wide	
etchLe leasur te Ke <b>RAC</b> k me requi uppro ter bu emen hen 1 es. In s will	earn: Relievin rement with A	g User Burdens in utomated Statisti steres Sketch: Adaptive Measure	Approxima ical Inference reserver	te work-wide moref lage1 of 16	X= 3 0 10
etchLe leasur leasur requi approv requi appr	earn: Relievin, rement with A	g User Burdens in utomated Statisti setenge Sketch: Adaptive Measure	Approxima ical Inference Preserver	te work-wide	
etchLe leasur Tegeno trate Ke rk nate rreptor saving ser bk ser bk ser bk ser bk ser ki ser ke ser k	earn: Relievin rement with A Elastic : Fek yangton Q Institu Tec	g User Burdens in utomated Statisti stowaat Sketch: Adaptive Measure	Approxima ical Inference Freedorts	te work-wide	
etchLe leasur in the second requires the secon	earn: Relievin rement with A Elastic : Fek yangton Mathu Tec haan	g User Burdens in utomated Statisti Sketch: Adaptive Measure	Approxima ical Inference Freedorts and Fast Net ements counterfast	te work-wide	Z: C 0 0.00
etchLo leasur rete Kr RAC change rete Kr RAC change rete kr rete ser ba ser sta ser ser ser ba ser ser ser ser ser ser ser ser ser ser ser ser ser ser ser ser	earn: Relievin, rement with A Elastic : Pet Vangton Mastiru Tec haar A misorii	g User Burdens in utomated Statisti Sketch: Adaptive Measure	Approxima ical Inference Freedorts and Fast Net ements _ spention	te work-wide marafilage 1 of 10	All: with UnivM
etchLed Neasure Territoria Territ	earn: Relievin, ement with A Elastic : Fek Vangton Pek Minisoru ABTREACT	g User Burdens in utomated Statisti Sketch: Adaptive Measure Rethinking N Zaoring Lut. Antonie	Approxima ical Inference Figure 18 and Fast Net ements _second	te work-wide maret lage 1 of 10 Rule Them Monitoring	All: with UnivM ar. Vladimi Braverni
etchLo teacure are k reading and a second reading a se	earn: Relievin, ement with A Elastic : Fek Vangton Pek Minitiu Fek Minitiu Thyn netwoli Aminoni Attraction Minitiu Tuban etwoli Minitiu Miniti	g User Burdens in utomated Statisti Sketch: Adaptive Measure Rethinking N Zaoxing Liu <sup>r</sup> , Antonis	Approxima ical Inference Figure 18 and Fast Net ements are Sketch to letwork Flow	te work-wide mare at lange 1 of 10. Rule Them Monitoring orsanger', Vyas Sek - Carnegie Mellon U	All: with UnivM ar. Vladimir Braverm niversity
etchLed deasur ate Kr CRAM deasur ate Kr CRAM deasure ate Kr CRAM deasure ate Kr CRAM deasure ate Kr CRAM deasure ate Kr CRAM deasure ate Kr CRAM deasure ate Kr CRAM deasure ate Kr CRAM deasure deas	earn: Relievin, ement with A Elastic : Fek yangton Guittiu Thur Nation ABTRACT Wan stack, Januar Abtractor Holton at tack, Januar Martine Mart	g User Burdens in utomated Statisti stomated Statisti Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive	Approximatical Inference Freedorts and Fast Net ements ane Sketch to letwork Flow	te work-wide maret lage 1 of 10 Rule Them Monitoring orsanger', Vyas Sek - Carnegie Mellon U	All: with UnivM ar. Vladimir Braverm niversity
etchLed deasur	earn: Relievin, ement with A Elastic : Fek yangton Q Institu Ten Non attack Manana ABTRACT Wan steven San attack Dan Holton var du manor attack Dan Holton var du Manor Holton var du Holton	sternungerungerunger g User Burdens in utomated Statisti sternunger Sketch: Adaptive Measure Measure C Rethinking N Zaoxing Luir, Antonis ' Jor D Russen and Statisti	Annousier, Gregory V ne Sketch to letwork Flow	te work-wide manual lange 1 of 10 Turber hange deep orsanger, Vyas Sek - Carnegie Mellon U	All: with UnivM ar. Vladimir Braverm niversity
etchLcd Measur Text Text Construction Constr	Elastic : Elastic : Elastic : Elastic : Elastic : Elastic : Elastic : Elastic : Elastic :	stehnungstigger (1915) g User Burdens in utomated Statisti stormated Statisti stormated Statisti stormated Statisti stormated Statisti stormated Statisti Sketch: Adaptive Measure Measure Control Statistical Office Statistical Control Statistical	Approximation of the second se	te work-wide work-wide The second second Monitoring orsanger, Vyas Sek Carnegie Mellon U Dranger Mellon U Drange	All: with UnivM ar. Vladimir Braverm niversity
etchLdc deasur	Elastic : Elastic : Elastic : Frederica :	sterenegatigaper unter g User Burdens in utomated Statisti sutomated Statisti sutomated Statisti sutomated Statisti sutomated Statisti sutomated Statisti sutomated Statisti Sketch: Adaptive Measure Sketch: Adaptive Measure Statisti Stati	An Approxima ical Inference Farefult and Fast Net ements • security Manousis, Gregory V ms Hopkins University Manousis, Gregory V ms Hopkins University	te Second Second work-wide manual lange 1 of 10 Second Second Corrange Mellon U orsanger, Vyas Sek Carnegie Mellon U Second Second Second Corrange Mellon U Second Second Second Second Second Second Second Second Second Second Second Second Second Second Second Second Secon	All: with UnivM ar., Vladimir Braverm niversity
etchLdc deasur	Elastic : Elastic : Elastic : Fred Service	sterenegations in utomated Statisti utomated Statisti sterenegations Sketch: Adaptive Measure Sketch: Adaptive Measure Statisti Sketch: Adaptive Measure Statisti Sketch: Adaptive Measure Sketch: Adaptive Sketch: Adap	And Fast Net enand Fast Net enand Fast Net enand Fast Net enands science of the second science of the second s	te work-wide work-wide The second secon	All: with University is main factor and a revore a mile capacity of the second arc. Vladimir Braverm hversity
etchLed Measure The Measure of the M	Elastic : Elastic : Elastic : Free Service : Elastic : Elastic : Elastic : Free yangon Office : Per yangon I Minima : I Minima : Minima :	sterenegations in utomated Statisti utomated Statisti subornated Statisti subornated Statisti subornated Statisti subornated Statisti Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive Measure Statistics	Approximation and Fast Net encode and Fast Net encode sector and the sector sector and the sector and the sector and the sector and the sector and the sector and the sector and the sector and the sector and the sector and the sector	te work-wide work-wide termine the second seco	All: with University is main factor and yield requires a constraint of behavior and the const
etchLed Veasure The Article State of the Article St	Elastic: 1 Elastic: 1	g User Burdens in utomated Statisti sutomated Statisti sutomated Statisti sutomated Statisti sutomated Statisti sutomated Statisti Sketch: Adaptive Measure Sketch: Adaptive Measure Statisti Statisti Sketch: Adaptive Measure Statististi Statisti Statisti Statisti Statisti Statisti Statisti S	And Fast Net enand Fa	te work-wide Work-wide Work-wide Corsanger, Vyas Sel Carnegie Mellon U Corsanger, Vyas Sel Carnegie Mellon U Throma magnetic Mellon U Corsanger, Sanger Sel Sel Sol, et al. Sel Sel Sol,	All: with University Setting a setting of the setting of the setting of the setting of th
etchLed Veasure Transformer Read Transfo	earn: Relievin, ement with A Elastic :	televenupartypertertel g User Burdens in utomated Statisti د معنیه Sketch: Adaptive Measure Cara ا Cara ا Cara ا Cara ا Cara ا Cara ا Cara ا Cara I Cara I Car	Approximate and Fast Net energy and Fast Net energy sector sector backetch to letwork Flow Manousis, Gregory M manousis, G manousis, G m m m m m m m m m m m m m m m m m m m	te Contract of the second sec	All: while the second
etchLede teacher are the requirements are the requi	earn: Relievin, ement with A Elastic :	televangerategory of televang	And Fast Net end F	te Contract of the second sec	All: white and the second sec
etchLede teacher are the requirements are the requi	earn: Relievin ement with A	Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive Measure Cara a Cara a	And Fast Net end F	te Control of the second seco	All: which are a set of the set
etchLed Measure are Re Transfer are reason of the second of the second of the second of the second of the second of the second of the second of the second of the second of the second of the second of the second of the second o	earn: Relievin ement with A	toread and a second	And Fast Net end F	te Contract and the set of the s	All: white the second
etchLed Measure The Annual States and Ka Constant Constan	earn: Relievin ement with A Elastic: 1	Steetch: Adaptive Measure Steetch: Adaptive Measure Steetch: Adaptive Measure Control	Approximate     and Fast Net energy     and Fast Net energy     work     work     and Fast Net energy     work	te Contract of the second sec	A constraints of the constraints
etchLed deasure are Re- transfer are ready are	<section-header></section-header>	statistical production of the second productio	Approximate and Fast Net energy and Fast and Fast Net energy and Fast Net energy and Fast	te Contract of the second of	All: with University () All () with University () and () with University () with
etchLed Veasure Transfer Trans	<section-header></section-header>	stationary of the second secon	Approx. Approx. Approx. Approx. The Approx. The App	te Contract of the second of	All: with University. In mail caption of the second of th
etchLed Veasure Transferrence are by require the second require to the second requiret to the second requiret t	<section-header></section-header>	Sketch: Adaptive Measure Sketch: Adaptive Measure Sketch: Adaptive Measure Company Com	Approx. Ap	te Contract of the second sec	Artic and the second se



MARPLE [SIGCOMM'17]

SONATA [SIGCOMM'18]

Both papers enable operators to express monitoring queries

A compiler then compiles these queries to: switch programs + control code

The two papers differ among others in the types of queries they support



Develop techniques and tools to monitor *all flows* by

- relying on in-switch data structures (Bloom Filters) and
- decoding them at the controller-level



Develop P4-based detection mechanisms to

- diagnose TCP performance issue (e.g. small receiver buffers)
- heavy-hitter (e.g. port scanners, superspreader, DDoS)

Introduce techniques to make sketch-based monitoring more practical (by making sketches adaptive or "universal")



Data plane for programmability

Performance Monitoring Applications offloading

Platforms Correctness Management for Data plane programmability

#### [SOSR'15]

• • • • • • • • • • • • • • • • •	G Q Q	17pdf (page 1 of 7)	C Q D	road.pdf (page 1 of 14)
NetPaxos: Consensus at Network Speed	In-Network Compute Whose Time Amedeo Saniet, Ibrahim Al	ation is a Dumb Idea e Has Come bdelaziz Abdulla Aldilaijan.	SilkRoad: Making Stateful La Chean Using S	ayer-4 Load Balancing Fast and Switching ASICs
Hunder To Dears' Dearies Sciencis'	Marca Conini	Danas Kalnis	Cheap Using C	Witching ASICs
Huyini tu Dang Dahete Sciascia	Wateo Callin	UST	Rui Miao Hone	zvi Zeng Changhoon Kim
Marco Canini' Fernando Pedone' Robert Soulé	KA KA	031	University of Southern California Fac	cebook Barefoot Networks
<sup>*</sup> Università della Svizzera italiana <sup>†</sup> Université catholique de Louvain	ABSTRACT Programmable data plane hardware creates new opportuni- ties for infusing intelligence into the network. This raises a fundamental question: what kinds of computation should be	Programmable networks create the opportunity for in- network computation, i.e., offloading a set of compute opera- tions from end hosts into network devices such as switches and smart NICs. In-network computation can offer substan-	Jeongkeun Lee Barefoot Networks	Minlan Yu Yale University
ABSTRACT This paper explores the possibility of implementing the widely deployed Paxos consensus protocol in network devices. We present two different approaches: (i) a detailed design de- scription for implementing the full Paxos logic in SDN switches which identifies a sufficient set of required OpenFlow exten- which identifies a sufficient set of required OpenFlow exten- tion of the set of the	delegated to the network? In this paper, we discuss the opportunities and challenges for co-designing data center distributed systems with their network layer. We believe that the time has finally come for offloading part of their computation to execute in-network. However, in-network computation to as must be judiciously crafted to match the limitations of the network machine archi-	tial performance benefits, as it is for example the case with consensus protocols [9, 10] and in-network caches [20]. Al- though traditional networks are not capable of computation, the idea of using the network conclusion to the state of the top perform computation on transmitted data is reminiscent of Active Networks [30], which proposed to replace packets with small programs called "capables" that are executed at	In this paper, we show that up to hundreds of software load balancer (SLB) servers can be replaced by a single modern switching ASC, potentially reducing the cost of load balancing by over two orders of magnitude. Today, large data centers typically employ hundreds or thousands of servers to load-balance incoming traffic over application servers. These software load balancers (SLB) may packets destined to a	A. A newtrine torial. If an Mass, Incogen Zong, Changboon Kim, Joongkom Lee, and Min- Han Mass, Jong Zong, Changboon Kim, Josef Layer et al. Cond Baharong: The num of Chen USB Using Switching ASICs. In Proceedings of SIG- COMM '17, Zun Angeles, CA, IASA, August 21–25, 2017, 14 pages. https://doi.org/10.1145/3009822.3098824
sions; and (ii) an aitemative, optimistic protocol which can be implemented without changes to the QoenFlow APJ, be relies on assumptions about how the network onless mes- sages. Although neither of these protocols can be fully in- plemented without changes to the underlying switch firmware, we argue that such changes are feasible in existing hardware. Moreover, we present an evaluation that suggests that mov- ing Paxos logic in the network woll del significant per focuses specifically on the Paxos consensus	tecture of programmable devices. With the help of our exper- iments on machine learning and graph analytics workloads, we identify that aggregation functions raise opportunities to exploit the limited computation power of networking hard- ware to lessen network congestion and improve the overall application performance. Moreover, as a proof-of-concept, we propose DATET, a system that performs in-network data agreements. Theoremental results with an initial protetorue	each travened switch. However, for the past two decades the hardware capabilities were lacking. This appears to be changing. The recently proposed RMT architecture [6] and its upcom- ing incarnation in the Barefoot Networks' Tofino [3] switch chip has a flexible parser and a customizable match-ascion engine. To process packets at high speed, this architecture has a multi-stage pipeline where packets flow at line rate. Each	service (with a virtual IP addres, or VIP), to a pool of servers tasked with providing the service (with multiple direct IP addresses, or DIPs). An SLB is stateful, it must always map a connection to the same server, even if the pool of servers changes and/or if the load is spread differently across the pool. This property is called <i>pre-connection consistency</i> or PCC. The challenge is that the load balancer must keep track of millions of connections simultaneously.	1 INTRODUCTION Stateful layer-(1,4) load halancers scale out services hested in doud datacenters by mapping packets destined to a service with a virtual IP address (UPI) to a pool servers with multiple direct IP addresses (DIPs or DIP pool). I4 load halancing is a critical function for inbond traffic to the cloud and traffic across tenants. A previous study [39] reports that an average of 44% of cloud traffic is VIP traffic and thus
formance benefits for distributed applications. <b>Categories and Subject Descriptors</b> C.2.4 (Distributed Systems): Reviow operating systems: C.4 (Performance of Systems): Reliability: availability, and serviceability: D.4.5 (Reliability): Fault-tolerance	show a large data reduction ratio (86.94-89.3%) and a similar decrease in the workers' computation time. <b>1 INTRODUCTION</b> The advent of flexible networking hardware [6] and express-	stage has a fixed amount of time to process every packet, allowing for lookups in memory (SRAM and TCAM), manip- ulating packet metadata and stateful registers, and performing boolean and arithmetic operations using ALUs. Other ven- dors are also introducing new classes of programmable chips with similar capabilities [7]. We believe that with this new generation of lexible data plane hardware it is worth revis-	Until recently, it was not possible to implement a load balancer with PCG in a merchant switching ASIC, because high-performance switching ASICs typically can not maintain per-connection states with PCC. Never switching ASICs provide resources and primitives to enable PCC at a large scale. In this paper, we explore how to use switching ASICs to build much faster load balancers than have been built before. Our system, called SilkRoad, is defined in a 400 line P4	needs load balancing function. Building cloud-scale L4 load balancing faces two major challenges: Support full bisection traffic with low latency: Data centers have rapid growth in traffic cholbing every year in Facebook [11] and growing by 50 times in six years in Google [40]. While the community has made efforts to scale out L2/L3 virtual switching to match full bisection bandwidth for intra- dutacent raffic (or full mateme scance) for full-bound traffic
I nira, moving consensus iogic into network devices would require extending the OnenFlow API with functionality that	sive data plane programming languages [5, 29] have produced	iting a fundamental question: as networks become capable	program and when compiled to a state-of-the-art switching	[17, 30], one missing piece is scaling L4 load balancers to match
Key works Software-defined networking, Paxos, NetPaxos. Software-defined networking, Paxos, NetPaxos. Implementing Paxos in the network provides a different point in the designs state: and distifiest at of net-	networks that are deeply programmable. The functionality of networks can now be enriched without hardware changes while retaining the capability of processing packets at very	of Comparison, while knows of comparisons around networks perform? In this paper, we will consider this question in the scope of data context another however it is likely that data con-	ASIC, we show it can load-balance ten minion connections simultaneously at this rate.	the full bisection bandwidth of the underlying physical network. Load balancing is also a critical segment for the end-to-end performance of delay-sensitive applications [23] and for low bisection bits of the constraint of the production of the production of the performance of t
1. INTRODUCTION work requirements for protocol implementors. This paper presents two different approaches. (i) a detailed description	grammable network devices are paving the way for new ser-	ters will be early adopters of programmable networks and	<ul> <li>Networks → Programmable networks; Network manage-</li></ul>	Ensure per connection consistency (PCC) during frequent DIP nod changes. Data constraints account at the change
way networks are configured and run. In contrast to tradi- dia sufficient set of OpenFlow extensions needed to imple-	vices to better support data center applications [9, 18] and improve network monitoring [13, 16, 24–26].	quirements. On the one hand, in-network computations can	ment; Data center networks;	ing to handle failures, deploy new services, upgrade existing requires, and react to the traffic increase [24]. Each oper-
tional networks, in which forwarding devices have propri- etary control interfaces, SDNs generalize network devices	<sup>‡</sup> Amedeo Sapio is also with Politecnico di Torino.	reduce latency and/or increase throughput of certain opera-	Load balancing; Programmable switches	ational change can result in many DIP pool changes. For example when we upgrade a service we need to bring down
using a set of protocols defined by open standards, including most prominently the OpenFlow [24] protocol. This move towards standardization has led to increased "network rop. Although neither of these protocols can be fully imple-	Permission to make digital or hand copies of all or part of this work for personal or classroom use is granted without fee provided that copies are not male or distributed for mofit or commercial adjunctures and that copies bear	tions. Furthermore, it can help reducing network traffic, so as to alleviate congestion, which is a major cause of application performance degradation. In particular, a computation that	Permission to make digital or hard copies of all or part of this work.	DIPs and upgrade them one by one to avoid affecting the ser- vice capacity. Such frequent DIP pool updates are observed from the service data and the servic
Permission to make digital or hud copies of all or part of this work for prossible classroom are is parted without for powided that copies are to make or dashibted for politic commonical advanger at that copies bear that copies are the software and the copies bear to access the advanger at the copies bear that copies are the software and the copies bear that copies are the software and the copies bear that copies are the software and the copies bear that copies are the software and the software and the software and the copies are the software and the software and the copies are the software and the software and the copies are the software and the softwar	the solice and field clusters on the space. Copyright of the Molyter Wett this notice and the fill clusters on the large QC. Copyright here components of this work owned by other han the authors(s) must be homed. Abstracting with credit is premission from composition of the control of the premission from premission (Bornaux, ergendish, tap cost energy or to redshiftshift bids, require yof or specific premission and/or a fee. Repert premission from premission (Bornaux, Park), Park and Allon, CA, USA 0. 2017 Copyright hed by the ownercharbor's). Bublication afplies lessed to	happens on-path and at line rate is appealing since it bears no cost to the application, which can spare CPU cycles for other tasks instead. On the other hand, despite recent technological advancements, network devices have limited compute power and little storage to support general computation. Moreover, systems designers are prescribed by the end-to-end princi-	ior personal or canserous use any grantex without he provided that and that copies hear this notice and the full station can the first page. Copyrights for components of this work owned by others than ACM must be honorea. Altestating with a cored is permitted. To copy otherwise, or republish, to post on servers or to relativistic to takes, requires prior specific permission and post as for the permission of the state of the server of the server of the server of the SIGCOMM '17, August 21:-35, 2017, Los Augusta, CA, USA 6 2017 Association for Couputing Machinery.	rom a ange web service provider with about a hundred of data center clasters (§3.1). During a DIP pool change, it is critical to ensure per connection consistency (PCC), which means all the packets of a connection should be delivered to the same DIP. Sending packets of an ongoing connection to a different DIP breaks the connection. It often takes subseconds to seconds for the connection.
Copylph is bill by the environments I, Nelscon agins licensed to ACM. ACM '99-14405 3551 34768-3551 350. http://dx.doi.org/10.11457712093.372809.	Association for Computing Machinery. ACM ISBN 978-1-430-556-98/1711515.00 https://doi.org/10.1145/3152434.3152461	pre [25] to avoid implementing application-specific logic in	ACM ISBN 978-1-4503-4653-5/17/08815.60 https://doi.org/10.1145/3098822.3098824	apparations to recover from a proken connection (e.g., one second in Wget), which significantly affects user experience.

[HotNets'17]

Consensus at network speed

#### In-Network Aggregation

Stateful layer-4 load balancers

[SIGCOMM'17]

(e.g., for MapReduce, graph analytics, ML)

+ NetCache [SOSP'17], NetChain [NSDI'18]

#### [SOSR'15]

	deiet.hotnets12.pdf (page	e 1 ef 7)	sigcommt7-sikros	d pdf (page 1 of 14) د م ال م ا
NetPaxos: Consensus at Network Speed	In-Network Computatio Whose Time Ha Amedeo Sapio <sup>4</sup> , Ibrahim Abdelaz Marco Canini, Panc	on is a Dumb Idea as Come ziz, Abdulla Aldilaijan, os Kalnis	SilkRoad: Making Stateful Lay Cheap Using St	er-4 Load Balancing Fast and vitching ASICs
Marco Canini <sup>†</sup> Fernando Pedone <sup>*</sup> Robert Soulé <sup>*</sup>	KAUST		Rui Miao Holigy University of Southern California Food	Zeng Changhoon Kim
<sup>*</sup> Università della Svizzera italiana <sup>†</sup> Université catholique de Louvain	ABSTRACT Pr Programmable data plane hardware creates new opportuni- ties for infusing intelligence into the network. This raises a fundamental question: what kinds of computation should be	rogrammable networks create the opportunity for in- work computation, i.e., offloading a set of compute opera- s from end hosts into network devices such as switches smart NCs. In-network computation can offer substan-	University of Southern Canfornia racet Jeongkeun Lee Barefoot Networks	Minlan Yu Yale University
<ul> <li>ASTRACE</li> <li>This paper acplores the possibility of implementing the why deployed Passo consensus protocols (in a detailed design deployed Passo sonsensus to demonstrate and changes to the Deeploy deployed protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield significant protocols (in the design space) (in the network would yield space) (in the design space) (in the</li></ul>	fundamental question: what kinds of computation should be delegated to the network? In this paper, we discuss the opportunities and challenges for co-designing data center distributed systems with their network layer. We believe that the time has finally come for offloading part of their computation to execute in-network. However, in-network computation to execute in-network, However, in-network computation takes and the initiations of the network machine archi- tecture of programmable devices. With the help of our exper- iments on machine learning and graph analytics workloads, we identify that aggregation functions raise opportunities to exploit the limited computation power of networking hard ware to lessen network congestion and improve the overall in ga- gregation. Experimental receives as a proof-of-concept, we propose DAIET, a system that performs in-network dapa aggregation. Experimental receives with an initial prototype show a large data reduction ratio (86.99-80.396) and a similar decrease in the workers' comprutation time. <b>D INTERODUCTION</b> The advent of flexible networking hardware [6] and expre- sive data plane programmable. The functionality of networks can now be erriched without hardware changer while reclaiming the capability of processing packets at very high rates, even above Terrabits per second. Emerging pro- duct arease in the work recomprutation time.	smart NCs. In-network computation can offer substan- performance benefits, as it is for example the case with sensus protocols [9, 10] and in-network caches [20]. Al- ight traditional networks are not capable of computation, alea of using the network not just to move data, but also efform computation on transmitted data is reminiscent <i>citive Networks</i> [30], which proposed to replace packets is markened and called 'capable'. This are executed at inverse of a second second second second second second transformed second second second second second transformed second second second second second second second second secon	<text><text><text></text></text></text>	ACH Microsove format: Itsi Muss, Isong 27 Serg, Changhoon Kim, Joonghoon Lee, and Min- na Yu. 2017. SilkHoud: Making Stateful Layer 4 Load Balancing Set and Chen Dising Warking ASICs. In Proceedings of SIG- COMM '17. Los Angeles. CA. USA, Angest 21 – 25, 2017, 14 pages. http://dx.ucg/ith.1145/200822.2300821 21 March 2019. A state of the state of the state of the state in cloud datacenters by mapping packets destined to a service with a virtual Padress (VIP) to a pool of servers with multiple direct IP addresses (DIP's or DIP pool). L4 load landning its acritical function for inhound traffic to the doal and traffic across transats. A previous study [20] reports that verselvs bad balancing function. Bulking chouse-scale L4 load and traffic across transats. A previous study [20] reports that verselvs bad balancing function. Bulking chouse-scale L4 load and traffic across transats of service the strain of the service study [20] reports the traffic doubling every scale fully. How the community has made efforts to scale out L2/L3 Virtual wirkding prevision bandfully in trans- tate transation prevision balancing the cites balancing the strain (17, 20), one missing prevision balancing the other Landbalancing is a critical angement for the end-to-end landbalancing is also a critical angement for the end-to-end competition of the strain strain of the prevision balancing for prevision Landbalancing is also a critical angement for the end-to-end performance of delaysensitive applications [23] and for box Landbalancing is also a critical angement for the end-to-end performance of delaysensitive applications [23] and for box Landbalancing is also a critical angement for the end-to-end performance of delaysensitive applications [23] and for box Landbalancing is also a critical angement for the end-to-end performance of delaysensitive applications [23] and for box Landbalancing is also a critical angement for the end-to-end performance of delaysensitive applications [23] and for box Landbalanc
Software-defined networking (SDN) is transforming the of a sufficient set of OpenPlux extension needed to imple-	vices to better support data center applications [9, 18] and many	ny of these applications have stringent performance re-	ment; Data center networks;	DIP pool changes: Data center networks are constantly chang-
way networks are configured and run. In contrast to tradi- tional networks, in which forwarding devices have propi- etary control interfaces. SDNs generalize network devices using as et of protocols defined by open standards, including most prominently the OpenFlow [24] protocol. This move towards standardization has led to increased "network pro- Permission on the digited hand copies are for the work proposed dearoom use is guard whooly for provided the topies are to make digited hand copies and the opies hand the opies are to make digited hand copies and the opies hand the opies are to make digited hand copies and the opies hand the opies are to make digited hand the opies are to make the difficult of the makes the following contributions:	improve network monitoring [13, 16, 24–26], equiri and the state of t	ements. On the one hand, in-network computations can anodly useful in several performance-oriented contexts to ace latency and/or increase throughput of certain opera- s. Furthermore, it can help reducing the trevork traffic, so as lleviate congestion, which is a major cause of application formance degradation. In particular, a computation that perso expath and all line rate is appealing since it bears no to the application, which is an spare CPU cycles for other sisted. On the other hand, despite recent technological ancements, network devices have limited compute power litel storage to support general computation. Moreover, ems designers are prescribed by the end-to-end princi- [28] to avoid immenning anofficiation specific locitic in	KEYWORDS Load balancing: Programmable switches Permission to make digital or hard copies of all or part of this work for personal or classroom use is granted without for powerlad hard of alterhistic for profile accounterial absundance page. Copyrights for components of this work owned by others than ACM must be honored. Alteracting with redit is permitted. To copy otherwise, or regulation, the profile of the individual to be for permissionstance. SIGCOMM '17, August 21–25, 2017, Los Angeles, CA, USA CAT Association for Comparing Machinery.	ing to and/m numes, depoy new services, upgrade existing services, and react to the traffic increase [24]. Each oper- ational change can result in many DIP pool changes. For example, when we upgrade a service, we need to bring down DIPs and upgrade them one by one to avoid affecting the ser- vice capacity. Such frequent DIP pool updates are doserved from a large web service provider with about a hundred of data centre clusters (§3.1). During a DIP pool change, it is critical to ensure per connection consistency (PCC), which means all the packets of a connections should be delivered to the same DIP. Sending packets of an ongoing connection to a different DIP breaks the connection. It often takes subseconds to seconds for applications to recover from a broken connection (c.e., one
And wells, and a structure of the struct	Association for Computing Machinery. ple [ ACM 1889 774-1493-596-397171131500 https://doi.org/10.1145/3152434.3152461	[28] to avoid implementing application-specific logic in	ACM ISBN 978-1-4203-4633-5/17/08. 815.00 https://doi.org/10.1145/3098822.3098824	applications to recover from a broken connection (e.g., one second in Wget), which significantly affects user experience.

[HotNets'17]

Consensus at network speed

#### In-Network Aggregation

Stateful layer-4 load balancers

[SIGCOMM'17]

(e.g., for MapReduce, graph analytics, ML)

+ NetCache [SOSP'17], NetChain [NSDI'18]

# NetCache: Balancing Key-Value Stores with Fast In-Network Caching

Xin Jin, Xiaozhou Li, Haoyu Zhang, Robert Soulé Jeongkeun Lee, Nate Foster, Changhoon Kim, Ion Stoica



NetCache solves the problem of load-balancing in key-values stores observing *dynamic*, *skewed* workload



NetCache solves the problem of load-balancing in key-values stores observing *dynamic*, *skewed* workload



It leverages that a small but very fast cache can provide perfect load-balancing... in theory



NetCache relies on the O(billion) throughput of programmable network devices to achieve it in practice



It relies on a tailored UDP-based protocol, an de/encoding scheme for storing variable length values, and sketches









Match	pkt.key == A	pkt.key =	== B	
Action	process_array(0)	process_	array(1)	
<u>pkt.value</u> :	A	B		
action p if pkt pkt. elif p arra	<pre>&gt;rocess_array(idx) :.op == read: .value</pre>	: .dx] odate: .ue	A B Registe	er Array













### Cache insertion and eviction

- Challenge: cache the hottest  $O(N \log N)$  items with **limited insertion rate**
- □ Goal: react quickly and effectively to workload changes with **minimal updates**




Data plane for programmability

Performance Monitoring Applications offloading

Platforms Correctness

for Data plane programmability

Management

"Data-plane" programmability goes beyond switch programmability (or P4 for that matter)

#### Offloading...

#### ... to FPGA-based SmartNICS

#### host networking

••• •••

age, and petabytes of formance, both ban cloud workloads, (

🚡 nsdit8-firestone.pd	f (page 2 of 15)
Azure Accelerated Networking: 5	SmartNICs in the Public Cloud
Daniel Firestone Andrew Putnam Sambhran	a Mundkur Derek Chiou Alireza Dabagh
Mike Andrewartha Hari Angenat Vivek	Bhanu Adrian Caulfield Eric Chung
Harish Kumar Chandrappa Somesh Chaturmohta	Matt Humphrey Jack Lavier Norman Lam
Fengfen Liu Kalin Ovtcharov Jitu Padhye G	autham Popuri Shachar Raindel Tejas Sapre
Mark Shaw Gabriel Silva Madhan Sivakumar Nis	heeth Srivastava Anshuman Verma Qasim Zuhair
Deepak Bansal Doug Burger Kushagra V	aid David A. Maltz Albert Greenberg
Micro	soft
Abtract domen cload anchitectures rely on each olicies such as tam- interostring anchitectures rely on each olicies such as tam- georgenergy of the second second second second second georgenergy of the second second second second second the second second second second second second second second georgenergy of the second second second second second second georgenergy of the second secon	all virtual networking features, such as private virtual net- works with customer supplied address spaces, scalable L4 load balancers, security groups and access control lists (ACLs), virtual outing tables, handwidth metering, QoS, and more. These features are the responsibility of the host platform, which bypically means software numing in the hyperboxic. The providing these services continues to in- crease, In the space of only a few years, we increased net- working speech by 40x and more, from IGRE to 40GnEs, and addec contracts, mainter that and while we built in- creasing capabilities, running this stack is notware on the host requires additional CPU sycles. Burning CPUs for these services takes way, from the processing power ani- tic service stack way, from the processing power ani- providing cloud services. Single Root IOV tranulization (SR-10V) [45] has been proposed to reduce CPU utilization by allowing direct ac- cess to NIC Indvance from the VM. However, this di- rect access would bypass the host SDN stack, making the NIC responsible for implementing all SDP policies. Since these policies change rupddy (weeks to nomba), we prominability while providing landware-like performance. In this paper we present Azure Accedented Network- ing (AcceNPet), rundus SDN stack, implemented on the FTGA-based Azure SmartNLC. AccelNet provides near- mative network performance in a virtualized environment, offloading packet processing from the host CPU to the Arear SmartNLC. AccelNet provides near- frantistic envirtues takes the profermance, of anitive network performance in a virtualized environment, offloading packet processing from the host CPU to the Arear SmartNLC. AccelNet provides near- frantistic envirtues takes the offload hydrome, with the programmability of software running in the hypervisor. Or goal is to present both our design and our experiences running AccelNet in production at scale, and lessons we kerned.
rmance, both bandwidth and latency, is critical to most	2 Background
oud workloads, especially interactive customer-facing	2.1 Traditional Host Network Processing
As a large public cloud provider, Azure has built its	environment such as the public cloud, all network I/O to

#### congestion control

#### HotCocoa: Hardware Congestion **Control Abstractions**

Monia Ghobadi Jennifer Rexford Microsoft Research Princeton University David Walker

Tahmasbi Arashloo Princeton University ABSTRACT

Mina

ontrol in multi-tenant data centers is an active ch because of its significant impact on customer and, consequently, on revenue. Therefore, new d protocols are expected to emerge as the Cloud lowing new consection control alevorithms in the equent updates, but pro vervisor and implementing rol algorithm, such as traf-oftware have well-studied ies. In this paper, we argue ontrol algorithm in ble NICs. To do so, we identify the absence of a simple high-level dona. as a broad set of congestion sedware implementation. It about low-level hardware primit ives. To HotCocca, we implement four congestion control ns (Reno, DCTCP, PCC, and TIMELY) and use sim-o show that HotCocca's implementation of Reno per-

s the behavior of a native imple

#### 1 INTRODUCTION

congestion control (CC) algorithms play a central role ta center network's efficiency and its tenants' quality erience. Hence, a significant number of congestion l algorithms concentrate on data center networks, which customizing their infrastructure to serve oads and tenants [3, 4, 8, 9, 23, 29, 31, 33], to continue, given the impact of network e and their rapid adoption

puting Machinery. 9-8/17/11...\$15.00

tors may deploy their CC algorithms in t While this approach enables frequen implementation, it incurs well-studied doing correction to the NIC (e.g., TCP Si and Generic Rec [2]). More recent to

f Reno per uch as Single Root I/0 NAT\_ACLs\_etc.) to the NICs [15]. We t

> even then, it is challenging and time hm. Thus, it takes nd deploy a new CC als

ingestion contact, and algorithms to enable higher-level abstractions that give operations and the second s structure across different CC definition of higher-level abstra enable th



[NSDI'18]

and from a physical d

[HotNets'17]

AMBRIDGE NetFPGA-SUME

## Host-based programmability + SmartNICs + programmable switches = fully programmable platforms

Big question is

#### How to combine them best?

#### ••• ••• • • beyond\_smart\_nics.pdf (page 1 of 6) 🗶 👻 📩 🛞 Q Search

#### **Beyond SmartNICs:** Towards a Fully Programmable Cloud

Adrian Caulfield Microsoft Research acaulfie@microsoft.com

(Invited Paper) Paolo Costa Microsoft Research pcosta@microsoft.com

Monia Ghobadi Microsoft Research mgh@microsoft.com

application to hardware and networks. In this paper, we focus on the potential of FPGA-based SmartNICs and programmable switches to realize this vision and illustrate some of the research challenges that need to be addressed to fully unleash its benefits.

I INTRODUCTION

However, the compute cycles-measured by the number of switches, e.g., [15], [16], this opens up exciting opportunities CPU cycles required to process each packet-are falling to rethink the way in which we design and deploy applications behind the massive acceleration in available network band-width [3]. As a result, CPU time is increasingly becoming a from the traditional strict boundary between network and contributor to the per-packet latency in high-speed cloud data application functions towards a fully programmable cloud, centers. To make matters more challenging, modern clouds in which application logic can be distributed across multiple are embracing a fully software-defined network (SDN) and accelerators and network devices. are increasingly expected to perform complex network policies such as regular expression matching and encryption [4]. Such applications that run in it. These benefits include application policies drive up the per-packet CPU cycles, which in turn specific control of network flows, the ability to run code increases the cloud costs and adds unpredictable latency to at precisely the right location in the network hierarchy, and the cloud services.

Traditionally, this has been addressed by offloading various a large-scale machine learning workload, a neural network networking functions to the Network Interface Cards (NICs) running on a distributed set of SmartNICs would benefit from such as TCP Segmentation Offload [5] and Generic Receive the direct interface to the network to reduce inference and Offload [6]. However, these techniques are not flexible enough training latencies. Further, programmable switches running to support complex policies. Techniques such as SR-IOV en- custom flow management code can reduce latency and optiable VMs to bypass the hypervisor and send packets directly to mize bandwidth by scheduling flows in an application-specific the NIC [7]. PCIe Process Address Space ID (PASID) reduces way, improving efficiency. Finally, the switch could even host the hardware resource requirements of SR-IOV, enabling it a parameter server, directly performing aggregation of the to scale to support containers or even individual processes. training weights from the SmartNICs below it. But, these techniques bypass the hypervisor, making it hard to After summarizing the key technology underpinning Smartenforce SDN-like policies.

as a new platform that enables network operators a flexible however, requires solving a number of novel and exciting environment to offload complex network policies and maintain research questions, which we outline in Section IV.

Abstract--FPGA-based SmartNICs and programmable switches have been recently introduced to leverage hardware acceleration and custom pipelines inside the cloud infrastructure. These devices are capable of handling the per-packet processing needs at line rate, including load balancing, encapsulation, congestion management, and security. We argue, however, that the benefits provided by these new devices could extend beyond software-defined networking use cases and they prompt a shift hardware-software co-design across all layers, ranging from application to hardware and networks. In this paper, we focus platforms and programming languages around FPGA-based SmartNICs [4], [12]-[14].

Most prior work treats the FPGA and the NIC domains separately by either focusing on the FPGA capabilities as a generic device or focusing on the NIC functions. In this paper, we turn our attention into the combined domains and argue that an FPGA-based SmartNIC should be thought of as both The continuous growth of cloud applications [1] is driving a programmable FPGA-based accelerator and a networking a steady increase in network infrastructure's bandwidth [2]. device. When combined with recently proposed programmable

A fully programmable cloud provides significant benefits to direct, low latency access to the network. For example, in

NICs in Section II, we describe our vision underlying the fully Recently, FPGA-based SmartNICs have been introduced programmable cloud in Section III. Implementing this vision,

IEEE International Conference on High Performance Switching and Routing, 2018 Data plane for programmability

Performance Monitoring Applications offloading

Platforms Correctness for Data plane programmability

Management

### So you've a programmable networks...

#### How do you make sure that it works as it should?!

p4v: Practical Verification for	r Programmable Data Planes*	Debugging P4 programs with Vera Radu Stoenescu Dragos Dumitrescu Matei Popovici Lorina Negreanu Costin Raiciu University Politchnica of Bucharest	[SIGCOM
Jeh Lin Barefoot NetworksWilliam Hullahan Lab University Net Haven, CT, UABChe Sch Barefoot State CherRobert Soulé Lagano, straterationHan Wang Barefoot Networks State CherChi Dat Barefoot Networks 	<text><text><text><text><text><text><text></text></text></text></text></text></text></text>	<text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text><text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text></text>	<text><text><text><text><text><text><text></text></text></text></text></text></text></text>

### So you've a programmable networks...

#### How do you make sure that it works as it should?!

p4v: Practical Verification for	Programmable Data Planes*	Debuggi Radu Stoenescu Drag	ng P4 programs with Vera os Dumitrescu Matei Popovici Lorina Negrea Costin Raiciu University Politebnica of Bucharest	
Jei Lin Bardon Nursch Bardon Nursch 	singer Milad Sharif Barefoot Networks Sunta Clara, CA, USA Sunford, CA, USA Nate Foster Cornell University Thaca, NY, USA Sunford, CA, USA Sunford, Sunford, CA, USA Sunford, Sunford, CA, USA Sunford, Sunford	<section-header><section-header><section-header><section-header><section-header><text><text><text><text><text><text></text></text></text></text></text></text></section-header></section-header></section-header></section-header></section-header>	<text><text><text><text><text><text><text><text><text><text><text><text><text></text></text></text></text></text></text></text></text></text></text></text></text></text>	<text><text><text><text><text><text><text><text></text></text></text></text></text></text></text></text>

## P4 by example

- P4 is a low-level language  $\rightarrow$  many gotchas
- Let's explore by example!
   IPv6 router w/ access control list (ACL)

```
control ingress { apply(acl); }
table acl {
  reads { ipv6.dstAddr: lpm; }
  actions { allow; deny; }
}
action allow() {
  modify_field(std_meta.egress_spec, 1);
}
action deny() { drop(); }
```

What could *possibly* go wrong?

#### What if we didn't receive an IPv6 packet?

ipv6 header will be invalid

#### What goes wrong

Table reads arbitrary values → Intended ACL policy violated

Can read values from a previous packet  $\rightarrow$  Side channel vulnerability!

Real programs are complicated: hard to keep validity in your head

```
control ingress { apply(acl); }
```

```
table acl {
   reads { ipv6.dstAddr: lpm; }
   actions { allow; deny; }
}
```

```
action allow() {
   modify_field(std_meta.egress_spec, 1);
}
```

```
action deny() { drop(); }
```

## **Property #1: header validity**

#### What if acl table misses (no rule matches)?

Forwarding decision is unspecified

#### What goes wrong

Forwarding behaviour depends on hardware

- May not do what you expect!
- Code not portable

```
control ingress { apply(acl); }
table acl {
  reads { ipv6.dstAddr: lpm; }
  actions { allow; deny; }
}
action allow() {
  modify_field(std_meta.egress_spec, 1);
```

```
action deny() { drop(); }
```

### Property #2: unambiguous forwarding

## **Types of properties**

#### **General safety**

- Header validity
- Arithmetic-overflow checking
- Index bounds checking (header stacks, registers, meters, ...)

#### Architectural

- Unambiguous forwarding
- Reparseability
- Mutual exclusion of headers
- Correct metadata usage (e.g., read-only metadata)

#### Program-specific

• Custom assertions in P4 program — e.g., IPv4 ttl correctly decremented

## Challenge #1: imprecise semantics



- P4 language spec doesn't give precise semantics
- Defined semantics by translation to GCL (a simple imperative language)
- Tested semantics
  - Symbolically executed GCL to generate input-output tests for several programs
  - Ran w/ Barefoot P4 compiler & Tofino simulator

## Challenge #2: modelling the control plane



## p4v overview • Automated tool for verifying P4 programs • Considers all paths • But also practical for **large programs** • Includes basic safety properties for any program • Extensible framework • Verify custom, program-specific properties • Assert-style debugging

Data plane for programmability

Performance Monitoring Applications offloading

Platforms Correctness

for Data plane programmability

Management

So you've a *verified* programmable networks... How do you manage it?!

How do you perform planned maintenance?

now that you've state in your switches...

How do you run multiple applications in your switches? monitoring, forwarding, load-balancing, etc.

How do you share resources amongst applications? especially memory and # packet operations

### We need an **Operating System** for the data plane

Definition Wikipedia

An operating system is a system software that manages computer hardware and software resources and provides common services for computer programs.

Do we have that? Nope. Not yet at least.

#### We're working on it...

#### [SOSR'17]



sis to compute which states require consistent migration and automatically augments the program to enable the transfer of these states at runtime. Our preliminary results indicate that Swing State is practical in migrating data-plane states at line rate with small overhead.

#### **CCS** Concepts

ond, even ignoring states dynamism, the exact set of states to be migrated may actually be unknown to the controller, preventing it from performing the migration in the first place. Indeed, the states location in memory can differ from device to device according to runtime factors (e.g. a hash computed on packet headers) that are invisible to the controller. Third, data-plane states



Source: Swing State: Consistent Updates for Stateful and Programmable Data Planes Luo et al., SOSR 2017

## Advanced Topics in Communication Networks





#### ~7 weeks

how to program in P4

>= 7 weeks in teams of 3

## Advanced Topics in Communication Networks





#### ~7 weeks

how to program in P4

>= 7 weeks in teams of 3

## The group project starts next week It accounts for 50% of your final grade

implementation

70%

achieves the basic goals is properly documented runs + results can be reproduced

implementation

70%

achieves the basic goals is properly documented runs + results can be reproduced

You'll have to write a detailed README (in Markdown) We'll provide you with a template

implementationachieves the basic goals70%is properly documented<br/>runs + results can be reproducedreportdescribes the main building blocks15%, 10 pages maxevaluates the solution<br/>describes what each group member did

implementation achieves the basic goals is properly documented runs + results can be reproduced describes the main building blocks report evaluates the solution describes what each group member did summarizes the problem and the solution presentation contains a *live* demo 15%, 10 min. +questions involves all group members

# The final deadline for the project is Wed Dec 16 at 23.59pm

This week	Select a proposal from the list (adv-net.ethz.ch) or send us your own proposal by email
<i>Every</i> week	Meet with the responsible assistant schedule a recurring slot in [10.15am; noon]
Mon Dec 16 11.59pm	Send us an archive with report, code, slides
Tue Dec 17 1.15pm—	Groups presentation + course/exam debrief attendance is mandatory

## The project has to be done in groups of 3 students "Matching" process for incomplete groups via Slack

Project grade is shared by each group member provided that each collaborated (roughly equally)

- Let us know in advance if that's *not* the case
- Briefly describe in the report the contribution of each group member
- Each group member should be involved in the presentation and be able to answer questions

## If you want to propose your own project, send us an email describing it by Thu Oct 31 11.59am

lvanbever@ethz.ch, cedgar@ethz.ch

### Quick overview of the proposals













Albert

Thomas

Roland

Alexander

Maria

Edgar

### Quick overview of the proposals



Albert Thomas Roland Alexander Maria Edgar

## Proposal #1 SDNSec: Forwarding Accountability for SDN Data Plane



Current data plane lacks accountability:

- Enforcing forwarding policies
- Validating that policies have not been violated
- Consistency guarantees under reconfiguration

## Proposal #1 SDNSec: Forwarding Accountability for SDN Data Plane



With SDNSec:

 Ingress-switch adds path in header

 Core-switches extract header, decrypt and forward

Controller verifies policy

[ICCCN 2016] NEC Corporation Japan, ETH Zurich (Perrig et. al.)

Proposal #2 Herding the Elephants: Detecting Network-Wide Heavy Hitters with Limited Resources



Separating elephant from mice is key in network management:

- Sampling is not accurate and results are delayed
- App-specific sketches limit network visibility

[Semantic scholar] Princeton, Walter Robert J. Harrison (Rexford et. al.)

## Proposal #2 Herding the Elephants: Detecting Network-Wide Heavy Hitters with Limited Resources



- Herd provides accuracy network wide
- Switches allocate resources based on flow type
- Switches notify controller when local heavy hitter
- Controller finds global heavy hitters

Extension: Network-Wide Heavy Hitter Detection with Commodity Switches

[Semantic scholar] Princeton, Walter Robert J. Harrison (Rexford et. al.)
# Retroactive Packet Sampling for Traffic Receipts



- Network nodes could cheat in monitoring
- Performing better for selected samples
- Delayed disclosure mechanism prevents it
- Estimates loss-rate and delay from controller

Extension: SQR: In-Network Packet Loss Recovery

[SIGMETRICS 2019] EPFL Lausanne, ETH Zurich (Perrig et. al.)

# Quick overview of the proposals



Albert Thomas Roland Alexander Maria Edgar

# Blink: Fast Connectivity Recovery Entirely in the Data Plane NSDI'19



# Blink: Fast Connectivity Recovery Entirely in the Data Plane NSDI'19



### Goal: improving blink

1. Selecting flows with low RTTs

2. Monitoring backup next-hops continuously to reroute faster

3. Monitoring the throughput to improve accuracy

NetCache: Balancing Key-Value Stores with Fast In-Network Caching SOSP'17 (for 2 students only)

Traditional way to implement a key value store:



NetCache: Balancing Key-Value Stores with Fast In-Network Caching SOSP'17 (for 2 students only)

Traditional way to implement a key value store:



### NetCache:



NetChain: Scale-Free sub-RTT CoordinationNSDI'18(for 2 students only)

Traditionally, key-value stores are replicated for *fault-tolerance* 



Coordination servers running a consensus protocol NetChain: Scale-Free sub-RTT CoordinationNSDI'18(for 2 students only)

Traditionally, key-value stores are replicated for *fault-tolerance* 





running a consensus protocol

# Quick overview of the proposals



Albert Thomas Roland Alexander Maria Edgar

NetHide: Secure and Practical Network Topology Obfuscation

### If I receive a packet to X with TTL = i, I should send it to Y with TTL = j



1

# pForest: In-Network Inference with Random Forests



iTAP: In-Network Traffic Analysis Prevention Using Software-Defined Networks



# Quick overview of the proposals



Albert Thomas Roland Alexander Maria Edgar

# Fast String Searching on PISA

P4 is very limited, e.g. it cannot work with strings. Or can it? It can even handle regular expressions!



Match		Action
state	chars	Action
0	do	<pre>set_state(1)</pre>
3	og	accept(4)
1	g <b>*</b>	accept(2)
0	*d	<pre>set_state(3)</pre>

\$ grep P4 \ lecture.txt

*In the control-plane:* Translate regex to automaton.

*In the data-plane:* Execute automaton using recirculation. Evaluation:

Compare to grep and co.

[SOSR 2019] USI, Barefoot (Jepsen et. al.)

# SilkRoad: Making Stateful Layer-4 Load Balancing Fast and Cheap Using Switching ASICs

*SilkRoad* using a P4 switch to replace software load balancers. It can handle millions of stateful connections using multi-level caching.





*In the control-plane:* Accept incoming connections. *In the data-plane:* Keep track of existing connections. Evaluation:

Test performance at large scale.

[SIGCOMM 2017] USC, Yale, Facebook, Barefoot (Miao et. al.)

# A Distributed Algorithm to Calculate Max-Min Fair Rates Without Per-Flow State

*s-Perc* is a congestion control algorithm that proactively assigns per-flow sending rates without per-flow state on devices.







*In the control-plane:* Implement the s-Perc algorithm.

*In the data-plane:* Create and parse control messages. Evaluation:

Compare with TCP and other protocols.

[SIGMETRICS 2019] Stanford University, MIT CSAIL (Jose et. al.)

# Proposal #7 Millions of Little Minions: Using Packets for Low Latency Network Programming and Visibility

In active networks, packets carry programs, which are run by switches.

Instruction		
LOAD, PUSH		
STORE, POP		
CSTORE		
CEXEC		





*In the control-plane:* Compile and start packet programs.

*In the data-plane:* Parse packets and execute instructions. Evaluation:

Test the performance of packet programs.

[SIGCOMM 2014] Stanford University, Cisco, Barefoot (Jeyakumar et. al.)

# Quick overview of the proposals



Albert Thomas Roland Alexander Maria Edgar

# **DIBS**: Just-in-time congestion mitigation for Data Centers



### currently

DC patterns can cause congestion.

Switches drop packets they cannot buffer.

with **DIBS** 

**\* detours** to neighboring switches.

minimizes drops, which speeds up job completion time.

# Marple: Language-Directed Hardware Design for Network Performance Monitoring



- The operator writes a query in a domain-specific language called Marple.
- \*The query is compiled into a switch program that runs on the network's programmable switches, augmented with new switch hardware primitives that we design in service of Marple.
- \*The switches stream results out to collection servers, where the operator can retrieve query results.

### 007: Democratically Finding the Cause of Packet Drops



Need to detect short-lived & concurrent failures despite noise

\*007 scales by uses traceroute to find paths of flows that had packet drops
\*007 finds faulty links democratically democracy by letting hosts vote

#### Implementation with p4 switches.

- \*detect retransmissions in switches
- **\***issue traceroutes directly from data plane
- \*combine traceroutes in control plane

# Quick overview of the proposals



Albert Thomas Roland Alexander Maria Edgar

#### Hardware-Accelerated Network Control Planes

[HotNets 2018] ETH, (Molero et. al.)

Modern programmable devices can perform small computations on billions of small packets per second. This paper shows how to leverage that to run control plane algorithms directly in the data plane



advertise vectors

### Seek and Push: Detecting Large Traffic Aggregates in the Dataplane

[arXiv 2018] CESNET, Cambridge (Kučera et. al.)

They present a data structure called *Elastic Tire* that is able to detect: heavy hitters, traffic shifts and superspreaders.



### Seek and Push: Detecting Large Traffic Aggregates in the Dataplane

[arXiv 2018] CESNET, Cambridge (Kučera et. al.)

They present a data structure called *Elastic Tire* that is able to detect: heavy hitters, traffic shifts and superspreaders.

High-level architecture:

- 1. Matching the flow using a dynamic LPM tree
- 2. Update Statistics
- 3. Control logic to update or report





#### Generic External Memory for Switch Data Planes

[HotNets 2018] CMU, Microsoft, Barefoot Networks (Kim et. al.)

Programmable switches are flexible but only have a limited on-ship SRAM and TCAMS



#### Generic External Memory for Switch Data Planes

[HotNets 2018] CMU, Microsoft, Barefoot Networks (Kim et. al.)

Programmable switches are flexible but only have a limited on-ship SRAM and TCAMS



Leverage RDMA to access remote memories at minimal latency and CPU usage



#### Generic External Memory for Switch Data Planes

[HotNets 2018] CMU, Microsoft, Barefoot Networks (Kim et. al.)

Packet buffer extension

Extending Lookup Tables

Extending State for network monitoring







# Advanced Topics in Communication Networks Programming Network Data Planes



Laurent Vanbever nsg.ee.ethz.ch

ETH Zürich Oct 29 2019

