



Extracting the Server from Serverless Computing! High-performance eBPF-based Event-driven, Shared-memory Processing

Shixiong Qi, Leslie Monis, Ziteng Zeng, Ian-chin Wang, K. K. Ramakrishnan

University of California, Riverside



Networked Systems Group

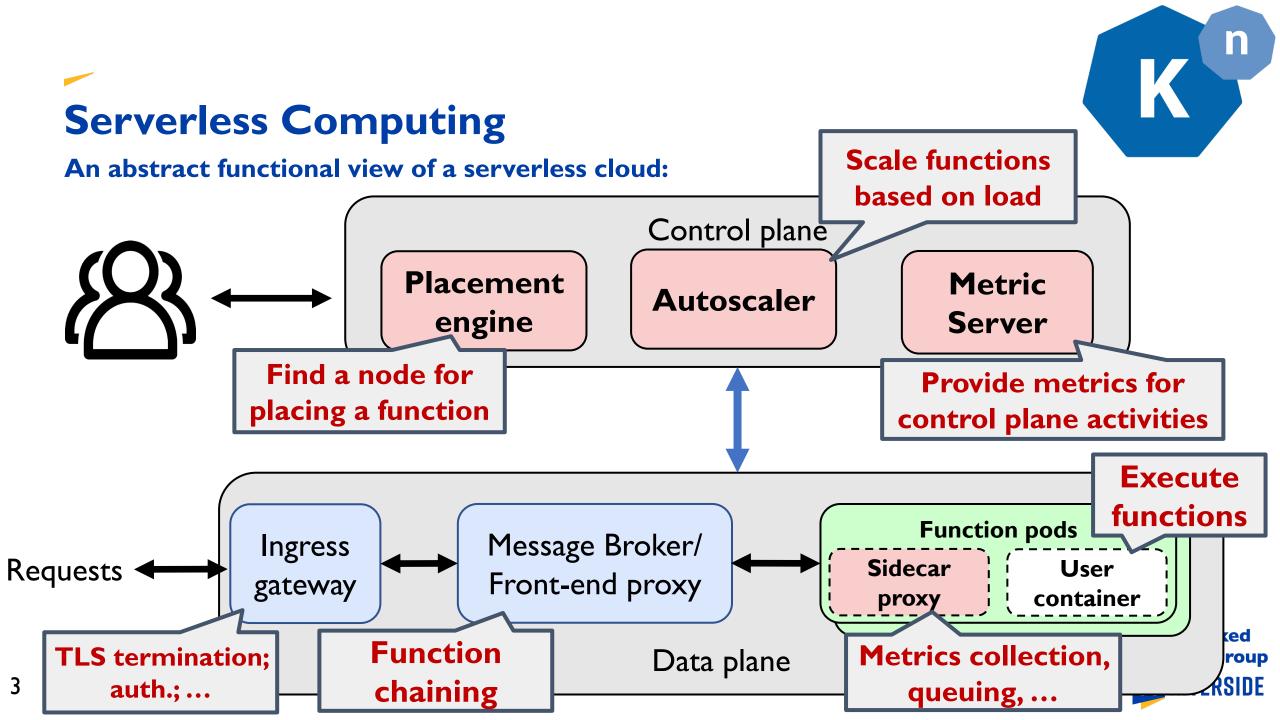
Visit us at: https://kknetsyslab.cs.ucr.edu/

Serverless Computing

What is Serverless computing?

- Paradigm for development and deployment of cloud applications to ease burden on users
 - Function as a service (FaaS): Users only provide application function code
 - Remove need for traditional always-on server components
 - Provisioning and managing the infrastructure becomes the cloud providers' job
- > Reduce user cost and complexity, and greatly improve service scalability and availability
- Challenges with serverless computing
 - Less focus on optimizing for high-performance, resource-efficiency, or being responsive
 - Need better support for both **low latency processing** and **low resource consumption**

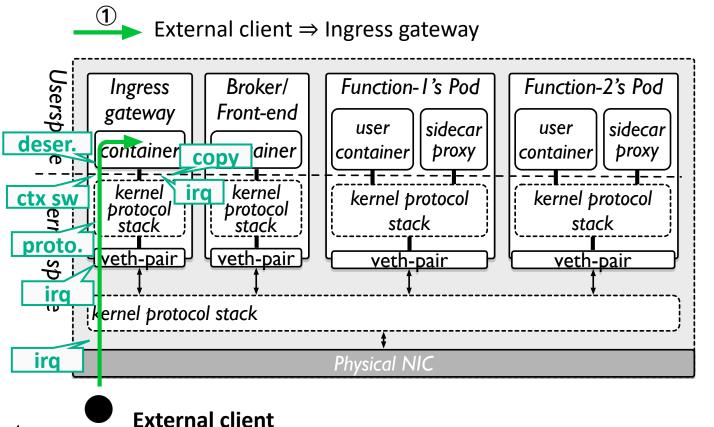
SPRIGHT: achieve high-performance, resource-efficient serverless function chains through shared memory and event-driven processing



Auditing the Overheads of Serverless Computing (1)

Processing involved in a typical serverless function chain setup: network protocol, copies, interrupts, context switches etc. abound

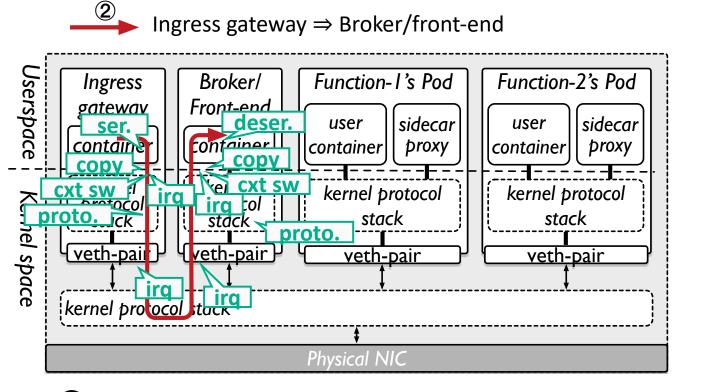
Ingress gateway: Intercept external requests; TLS Termination, authentication, etc



Data Pipeline		Exter	nal		With	in cha	in	Total
No.	1	2	total	3	4	5	total	Total
# of copies	1							
# of ctxt switches	1							
# of irqs	3							
# of proto. processing	1							
# of serialization	0							
# of deserialization	1							
								etwor

Auditing the Overheads of Serverless Computing (2)

Broker/front-end: an intermediate component for coordinating invocations within the function chain Broker: I copy, I context switches, 2 interrupts, I protocol processing, and I deserialization

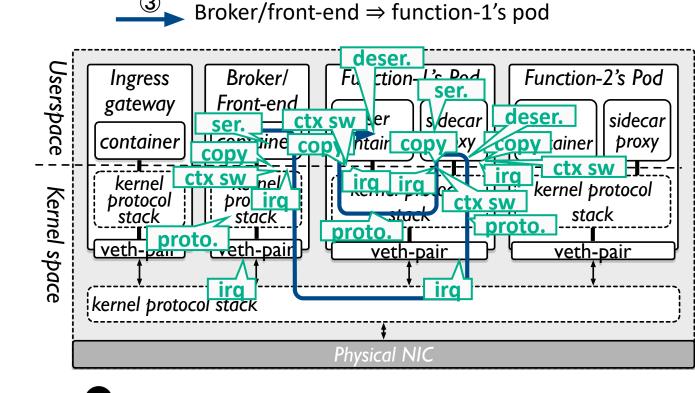


				vvicii	in cha	In	T - 1 - 1
1	2	total	3	4	5	total	Total
1	2	3					
1	2	3					
3	4	7					
1	2	3					
0	1	2					
1	1	1					etwor
	1 1 3 1 0	1 2 1 2 3 4 1 2 0 1	1 2 3 1 2 3 1 2 3 3 4 7 1 2 3 0 1 2	1 2 3 1 2 3 3 4 7 1 2 3 0 1 2	1 2 3	1 2 3	1 2 3



Auditing the Overheads of Serverless Computing (3)

Total: 4 copies, 4 context switches, 6 interrupts, 3 protocol processing, 2 serializations, and 2 deserializations **Sidecar**: 2 copies, 2 context switches, 2 interrupts, 1 protocol processing, 1 serialization, and 1 deserialization



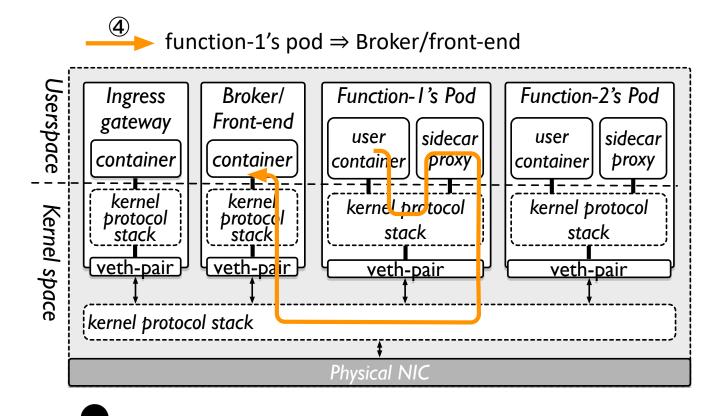
External client

6

Data Pipeline		Exter	nal		With	in cha	in	T I
No.	1	2	total	3	4	5	total	Total
# of copies	1	2	3	4				
# of ctxt switches	1	2	3	4				
# of irqs	3	4	7	6				
# of proto. processing	1	2	3	3				
# of serialization	0	1	2	2				
# of deserialization	1	1	1	2				

Auditing the Overheads of Serverless Computing (4)

Total: 4 copies, 4 context switches, 6 interrupts, 3 protocol processing, 2 serializations, and 2 deserializations **Sidecar**: 2 copies, 2 context switches, 2 interrupts, 1 protocol processing, 1 serialization, and 1 deserialization

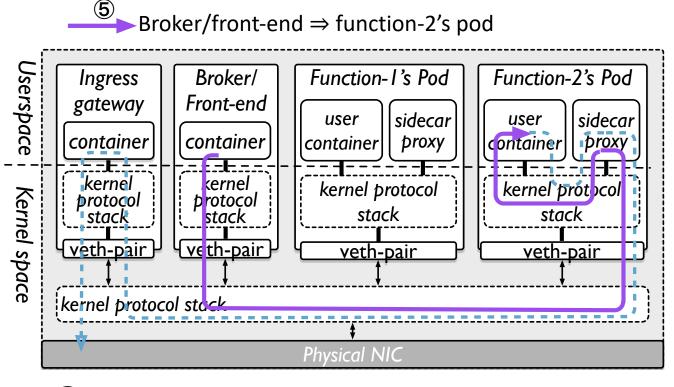


External client

Data Pipeline		Exter	nal		With	in cha	in	T 1
No.	1	2	total	3	4	5	total	Total
# of copies	1	2	3	4	4			
# of ctxt switches	1	2	3	4	4			
# of irqs	3	4	7	6	6			
# of proto. processing	1	2	3	3	3			
# of serialization	0	1	2	2	2			
# of deserialization	1	1	1	2	2			

Auditing the Overheads of Serverless Computing (5)

Total: 4 copies, 4 context switches, 6 interrupts, 3 protocol processing, 2 serializations, and 2 deserializations **Sidecar**: 2 copies, 2 context switches, 2 interrupts, 1 protocol processing, 1 serialization, and 1 deserialization



Data Pipeline		Exter	nal		With	in cha	in	Tetel
No.	1	2	total	3	4	5	total	Total
# of copies	1	2	3	4	4	4	12	15
# of ctxt switches	1	2	3	4	4	4	12	15
# of irqs	3	4	7	6	6	6	18	25
# of proto. processing	1	2	3	3	3	3	9	12
# of serialization	0	1	2	2	2	2	6	8
# of deserialization	1	1	1	2	2	2	6	7
								etwo stems

RSINF



*Processing in function chain complete; return a response

Auditing the Overheads of Serverless Computing

Key takeaways: Excessive overhead within the function chain

Takeaway#1: Excessive data copies, context switches, and interrupts.

Takeaway#2: Excessive, duplicate protocol processing for communication within the function chain

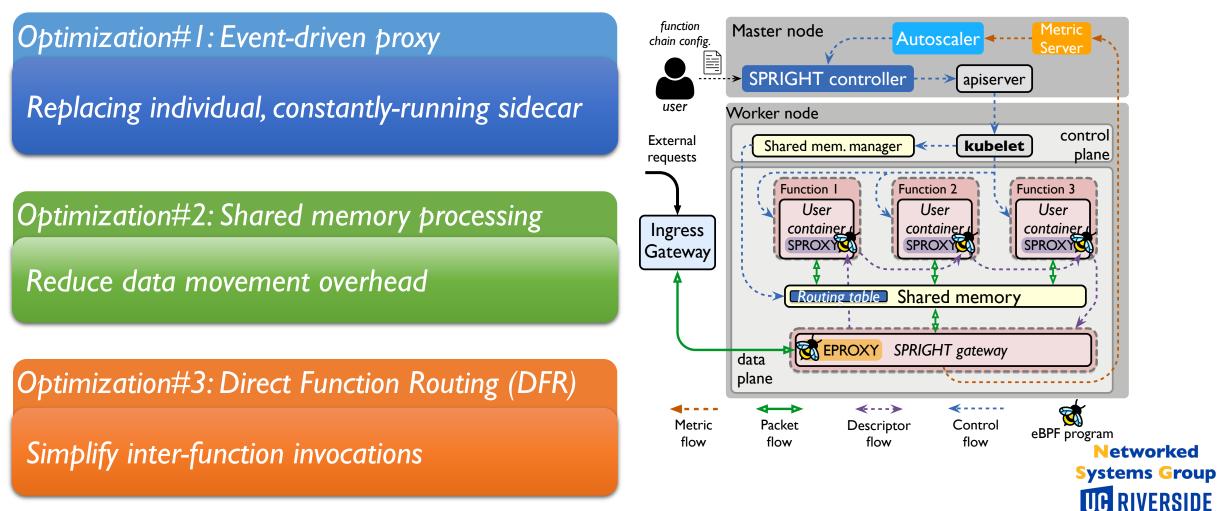
Takeaway#3: Unnecessary serialization/deserialization.

Takeaway#4: Individual, constantly-running components.

Data Pipeline		Exter	nal		With	in cha	in	Total
No.	1	2	total	3	4	5	total	Total
# of copies	1	2	3	4	4	4	12	15
# of ctxt switches	1	2	3	4	4	4	12	15
# of irqs	3	4	7	6	6	6	18	25
# of proto. processing	1	2	3	3	3	3	9	12
# of serialization	0	1	2	2	2	2	6	8
# of deserialization	1	1	1	2	2	2	6	7

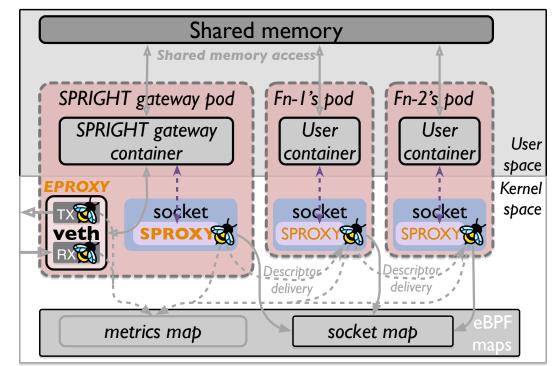
What is SPRIGHT?

eBPF-based event-driven capability + Shared memory processing



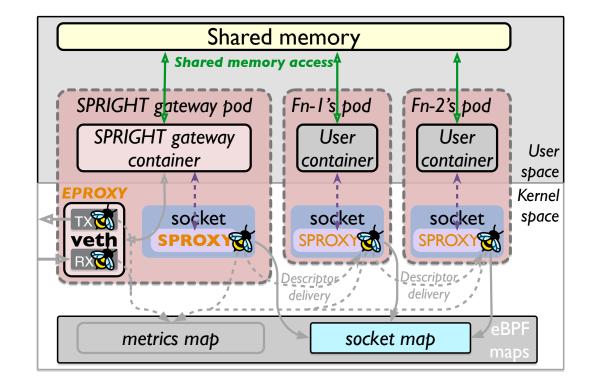
Optimization#I: eBPF-based Event-driven proxy (EPROXY and SPROXY)

- In-kernel eBPF-based "sidecar"
 - **EPROXY**: @Veth of SPRIGHT GW pod
 - Monitoring; iptables acceleration
 - **SPROXY**: Sidecar being injected at the socket level
 - Monitoring; Security; Routing
- Purely event-driven
 - No CPU overhead when there are no requests
- All in the kernel
 - Avoid extra user-kernel boundary crossings



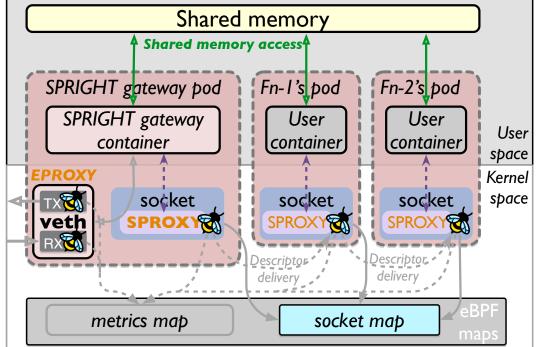
Optimization#2: Shared memory processing

- How to handle protocol processing?
 - SPRIGHT Gateway: Entry-point of a function chain
 - Consolidate kernel protocol processing
 - Move payload into shared memory



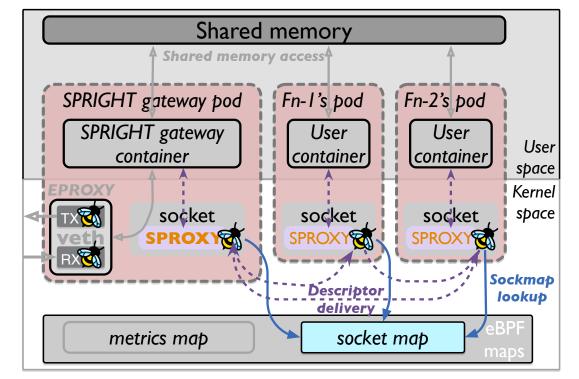
Optimization#2: Shared memory processing

- How to handle protocol processing?
 - SPRIGHT Gateway: Entry-point of a function chain
 - Consolidate kernel protocol processing
 - Move payload into shared memory
- Shared memory based data sharing between functions
 - NO copy, protocol processing, serialization, ...
 - Packet descriptor delivery: eBPF's socket message
 - reside in Event-driven proxy (**SPROXY**)
 - Socket-to-socket data transfer; Routing using eBPF's socket map
 - Strictly load-proportional compared to polling-based packet descriptor delivery (DPDK RTE Ring)



Optimization#3: Direct Function Routing

- Having the broker/front-end perform invocations between functions is unnecessary
 - Routing overhead
- DFR optimizes invocations within a function chain
 - The upstream function in the chain directly invokes the downstream function: bypass the gateway
 - DFR rules in eBPF's socket map
- DFR can reduce end-to-end latency of the function chain and improve the scalability
 - Eliminate an extra hop on the datapath
 - Benefit increases as the chain scales



Networked Systems Group

FRSINF

Overhead auditing: Knative vs. SPRIGHT

Data Pipeline		Extern	nal		With	in cha	in	Total
No.	1	2	total	3	4	5	total	Total
# of copies	1	2	3	4	4	4	12	15
# of ctxt switches	1	2	3	4	4	4	12	15
# of irqs	3	4	7	6	6	6	18	25
# of proto. processing	1	2	3	3	3	3	9	12
# of serialization	0	1	2	2	2	2	6	8
# of deserialization	1	1	1	2	2	2	6	7

• 0 data copies, 0 protocol processing, 0 serialization/deserialization overheads within the chain

The **event-based shared memory processing** brings substantial reduction of overheads for communication **within** the serverless function chain

etworked :ems Group RIVERSIDE [1] <u>https://github.com/GoogleCloudPlatform/microservices-demo</u>
 [2] <u>http://www.merl.com/wmd</u>
 [3] http://cnrpark.it/

Evaluation: across multiple serverless workloads

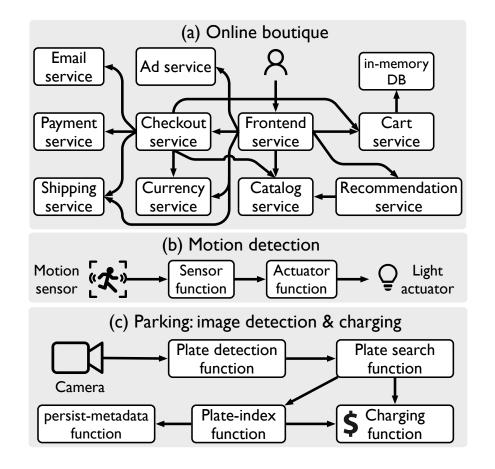
- I. Online Boutique from Google [I]
- Intense web traffic
- 10 functions
- <u>6 different sequences of function chains</u>

2. Motion detection [2]

- Intermittent IoT traffic (a burst every <u>few seconds</u>)
- 2 simple functions

3. Parking: image detection & charging [3]

- Intermittent & Periodic IoT traffic (once every <u>240</u> seconds)
- 5 functions with *heterogeneous* CPU service time for each (from 1ms to 435ms)



Performance with Online Boutique

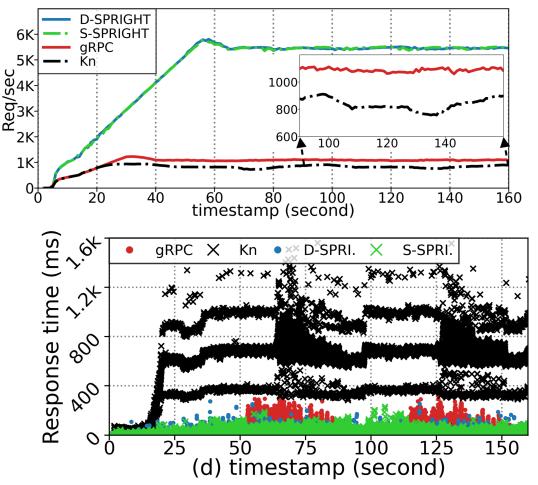
S-SPRIGHT vs. D-SPRIGHT vs. Knative mode vs. gRPC mode (no sidecar)

Throughput:

- Both D- and S-SPRIGHT maintain a stable RPS of ~5500 req/sec → (6× more than Knative)
- Without sidecar and front-end proxy, gRPC is slightly better than Knative, but its throughput is much lower than SPRIGHT (5x lower)

Latency:

- Knative shows clear overload behavior: high tail latency
- SPRIGHT's shared memory processing reduces communication overhead within function chains, achieving lower latency than Knative and gRPC, even at much higher traffic load



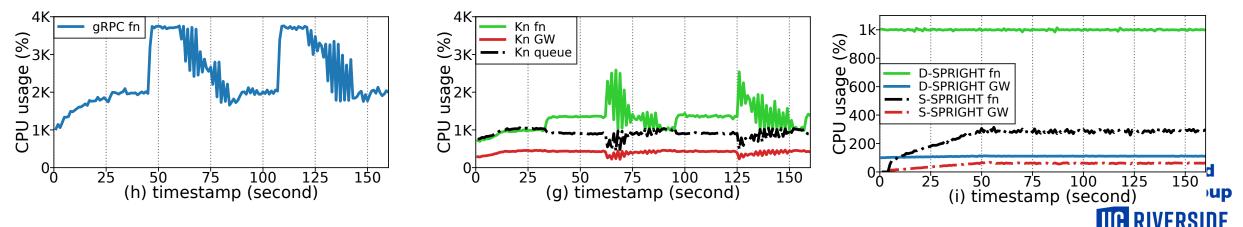
Note: for simplicity, we only show the latency results of a representative function chain in online boutique. More results can be found in the paper.

Performance with Online Boutique

S-SPRIGHT vs. D-SPRIGHT vs. Knative mode vs. gRPC mode (no sidecar)

Resource efficiency:

- entire Knative setup (including the gateway and queue proxies, which are constantly running): ~26
 CPU cores (out of 40)
- Entire gRPC setup (only functions, no gateway and sidecars): ~36 CPU cores (out of 40)
- D-SPRIGHT consumes **II cores** (one core for Gateway, **I0** cores for functions)
- S-SPRIGHT consumes in total only ~3 CPU cores

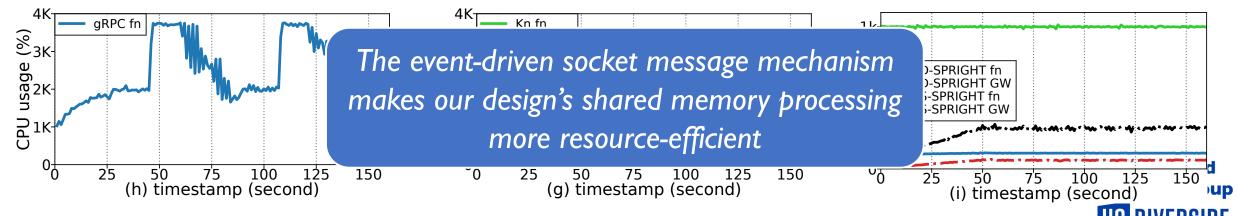


Performance with Online Boutique

S-SPRIGHT vs. D-SPRIGHT vs. Knative mode vs. gRPC mode (no sidecar)

Resource efficiency:

- entire Knative setup (including the gateway and queue proxies, which are constantly running): ~26
 CPU cores (out of 40)
- Entire gRPC setup (only functions, no gateway and sidecars): ~36 CPU cores (out of 40)
- D-SPRIGHT consumes **II cores** (one core for Gateway, **I0** cores for functions)
- S-SPRIGHT consumes in total only ~3 CPU cores



[1] <u>https://github.com/GoogleCloudPlatform/microservices-demo</u>
 [2] <u>http://www.merl.com/wmd</u>
 [3] http://cnrpark.it/



Evaluation: across multiple serverless workloads

I. Online Boutique from Google [I]

- Intense web traffic
- 10 functions
- <u>6 different sequences of function chains</u>

2. Motion detection [2]

- Intermittent IoT traffic (a burst every <u>few</u> <u>seconds</u>)
- 2 simple functions

3. Parking: image detection & charging [3]

- Intermittent & Periodic IoT traffic (once every <u>240</u> seconds)
- 5 functions with *heterogeneous* CPU service time for each (from Ims to 435ms)

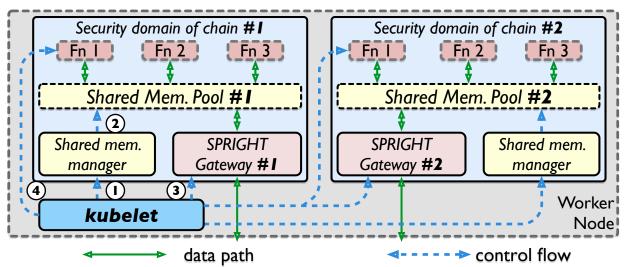
Our event-driven design sidesteps the need for cold start by keeping functions warm at minimum cost

Our design's event-driven features make it more efficient even if we keep functions warm compared to 'pre-warming' Knative functions

Shared memory considered harmful?

Our Solution: Security domain

- Trust model: functions within a chain trust each other, functions in different chains may not
- SPRIGHT constructs a security domain for each function chain with:
 - a private shared memory pool for each chain
 - DPDK's multi-process support
 - Use different shared data file prefix to separate memory pool
 - inter-function packet descriptor filtering with the **SPROXY**
 - Use **SPROXY** to construct packet descriptor filtering between functions
 - Restrict unauthorized access to the shared memory





Conclusion

22

SPRIGHT: event-driven + shared memory processing = load-proportional, high-performance

Using shared memory processing to optimize the data pipeline of current serverless function chains

Using event-driven processing to improve resource efficiency of current serverless function design

- When serving an intense web workload (online boutique):
 - **6x** throughput improvement, **70x** tail latency reduction and **30x** CPU usage savings over Knative
- When serving intermittent IoT workload (motion detection; parking):
 - Better than Knative even with 'pre-warmed' functions; side-stepping the 'cold-start'



Find SPRIGHT at: <u>https://github.com/ucr-serverless/spright</u>

