

#### **Piz Daint**

- 5704 XC50 nodes CPU + GPU, 64 GB memory.
- 1813 XC40 nodes CPU, 64/128 GB memory.







# **Tracking Wasted Money in HPC**







#### **Tracking Wasted Money in HPC**

#### Job Characteristics on Large-Scale Systems: Long-Term Analysis, Quantification, and Implications\*

Tirthak Patel Northeastern University Zhengchun Liu, Raj Kettimuthu Argonne National Laboratory

Paul Rich, William Allcock

Devesh Tiwari

#### A Case For Intra-rack Resource Disaggregation in HPC

GEORGE MICHELOGIANNAKIS, Lawrence Berkeley National Laboratory, USA

BENJAMIN KLENK, NVIDIA, USA

BRANDON COOK, Lawrence Berkeley National Laboratory, USA

MIN YEE TEH and MADELEINE GLICK, Columbia University, USA

LARRY DENNISON, NVIDIA, USA

KEREN BERGMAN, Columbia University, USA

JOHN SHALF, Lawrence Berkeley National Laboratory, USA

TACO, 2022

#### Quantifying Memory Underutilization in HPC Systems and Using it to Improve Performance via Architecture Support Gagandeep Panwar\* Da Zhang\* Yihan Pang\* Virginia Tech Virginia Tech Virginia Tech , Jeffrey Blacksburg, USA Blacksburg, USA Blacksburg, USA gpanwar@vt.edu daz3@vt.edu pyihan1@vt.edu Mai Dahshan Nathan DeBardeleben Binoy Ravindran Virginia Tech Los Alamos National Laboratory Virginia Tech nos, and Blacksburg, USA Blacksburg, USA Los Alamos, USA cations mdahshan@vt.edu ndebard@lanl.gov binoy@vt.edu Xun Jian Virginia Tech 2017 Blacksburg, USA MICRO, 2019 xunj@vt.edu

#### A Holistic View of Memory Utilization on HPC Systems: Current and Future Trends

and

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MEMSYS, 2021

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JSSPP, 2012







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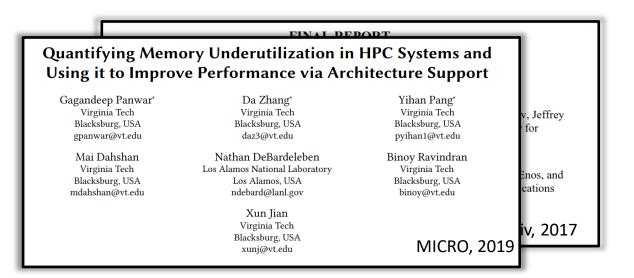
GEORGE MICHELOGIANNAKIS, Lawrence Berkeley National Laboratory, USA BENJAMIN KLENK, NVIDIA, USA BRANDON COOK, Lawrence Berkeley National Laboratory, USA MIN YEE TEH and MADELEINE GLICK, Columbia University, USA

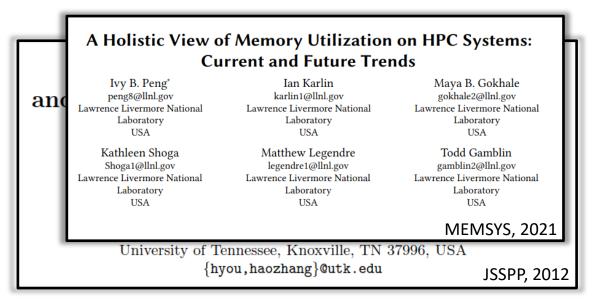
LARRY DENNISON, NVIDIA, USA

KEREN BERGMAN, Columbia University, USA

JOHN SHALF, Lawrence Berkeley National Laboratory, USA

TACO, 2022





Can we solve underutilization with sharing and fine-grained allocations?



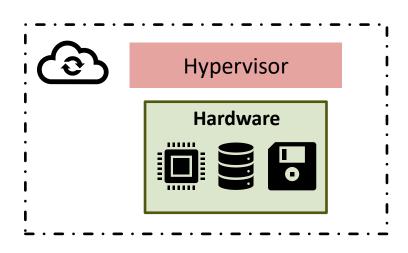
















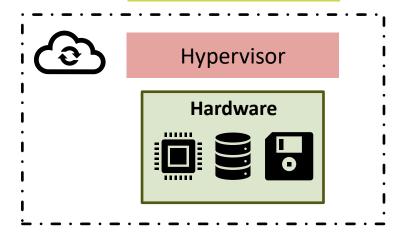


Function

**Application** 

Language Runtime

**Operating System** 







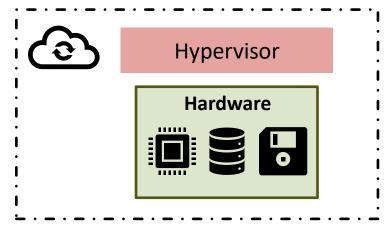


**Function** 

**Application** 

Language Runtime

**Operating System** 





**Virtual Machine** 





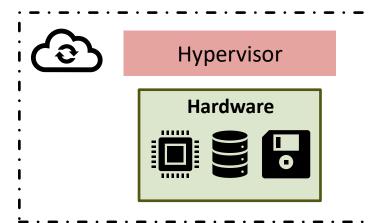


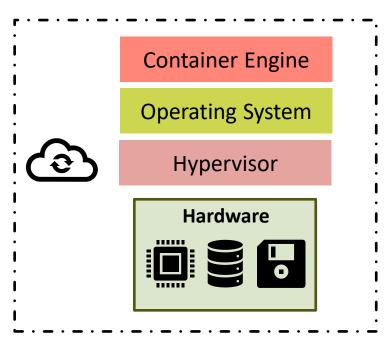
**Function** 

**Application** 

Language Runtime

**Operating System** 









**Containers** 







Function

**Application** 

Language Runtime

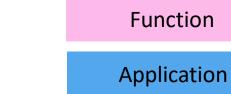
**Operating System** 

Hypervisor

Hardware







Language Runtime

**Container Engine** 

**Operating System** 

Hypervisor

Hardware









**Virtual Machine** 



**Containers** 







**Function** 

**Application** 

Language Runtime

**Operating System** 

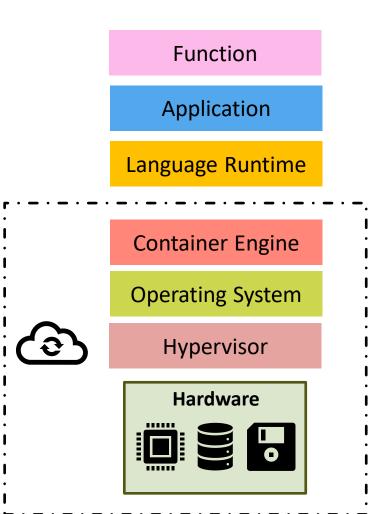
Hypervisor

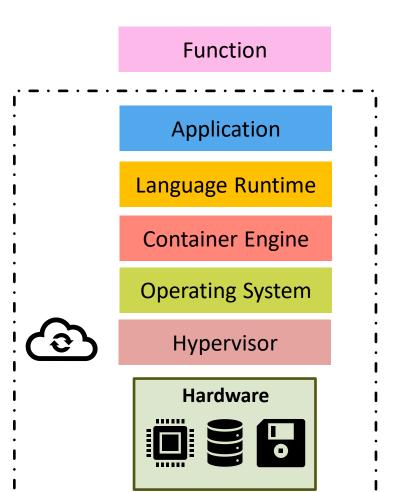
Hardware













**Virtual Machine** 



**Containers** 



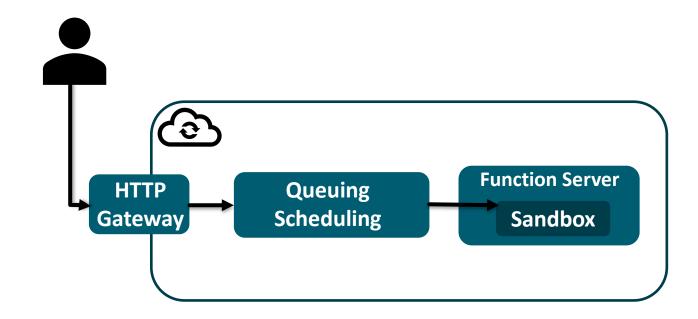
**Functions** 

**AWS Lambda** 







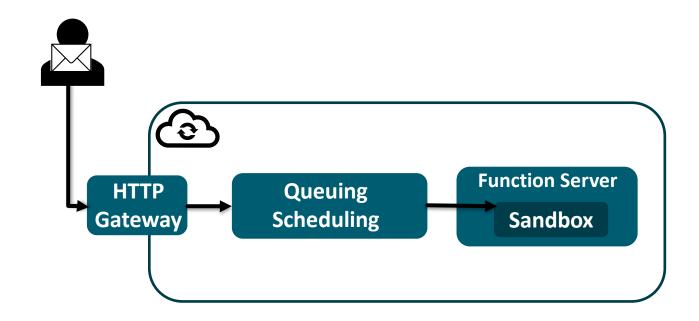










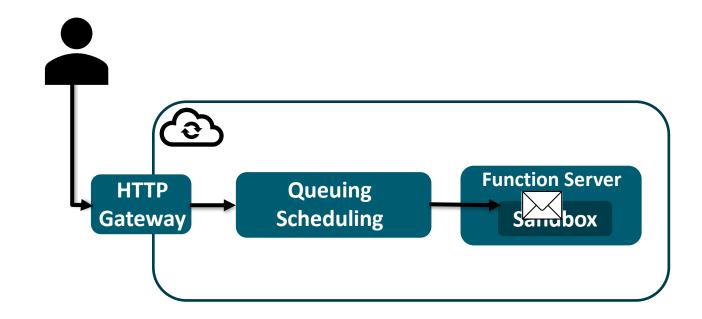










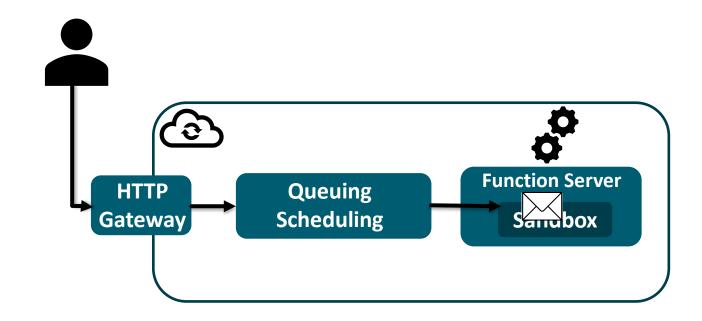










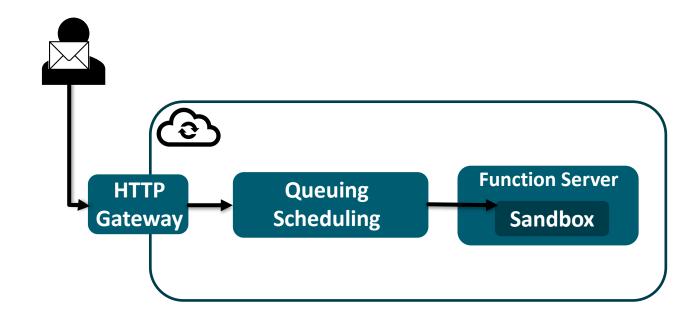




















# "But serverless is slow and expensive"







### "But serverless is slow and expensive"

# Scaling up the Prime Video audio/video monitoring service and reducing costs by 90%

The move from a distributed microservices architecture to a monolith application helped achieve higher scale, resilience, and reduce costs.



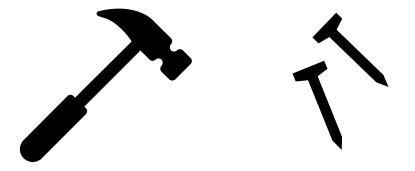




#### "But serverless is slow and expensive"

# Scaling up the Prime Video audio/video monitoring service and reducing costs by 90%

The move from a distributed microservices architecture to a monolith application helped achieve higher scale, resilience, and reduce costs.















Serverless is slow







Serverless is slow

Communication is slow and restricted







Serverless is slow

Communication is slow and restricted

Serverless is hard to program.







Serverless is slow

Answer: rFaaS

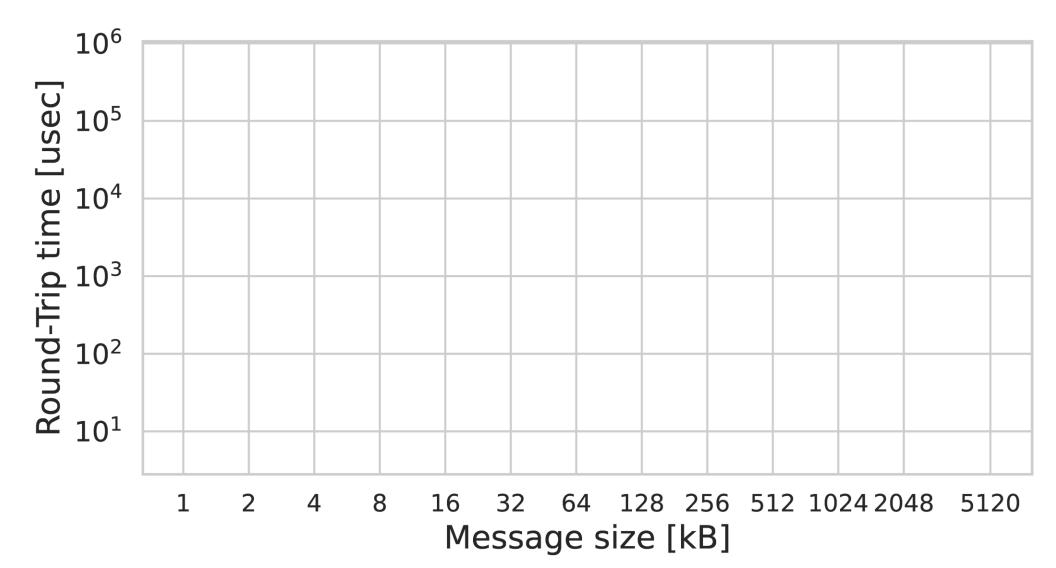
Communication is slow and restricted

Serverless is hard to program.







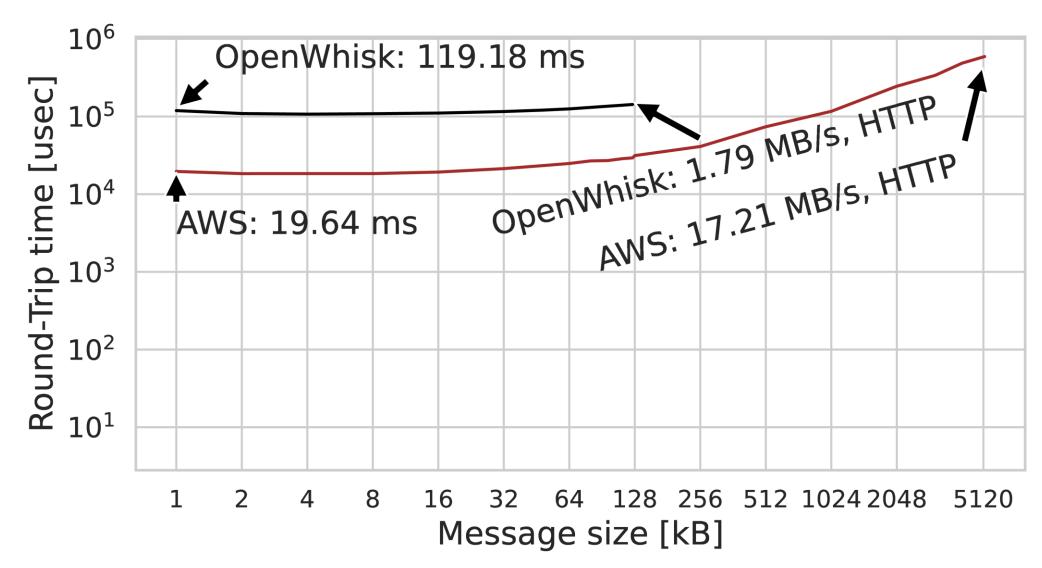










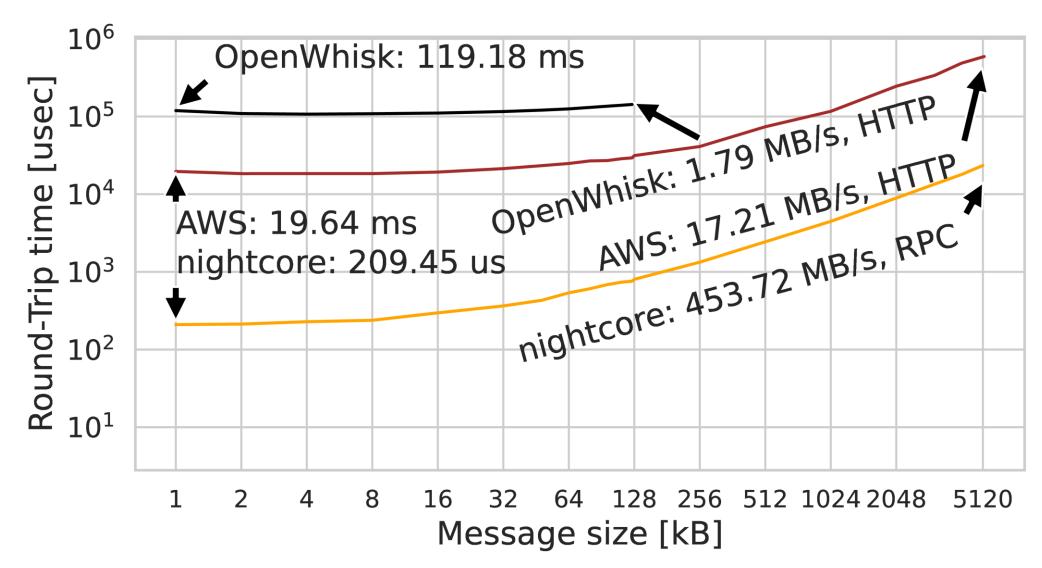












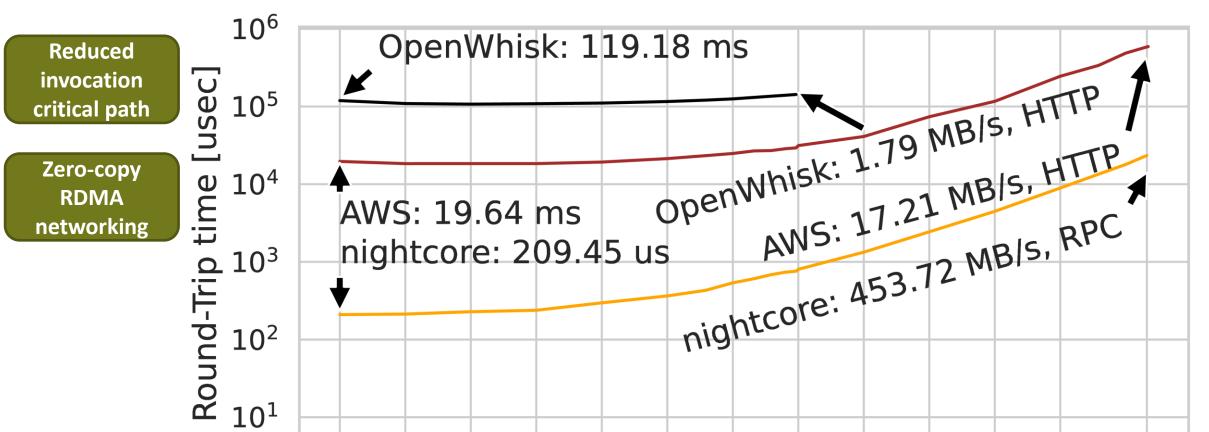








 $10^{1}$ 





128 256 512 1024 2048 16 5120 Message size [kB]

8

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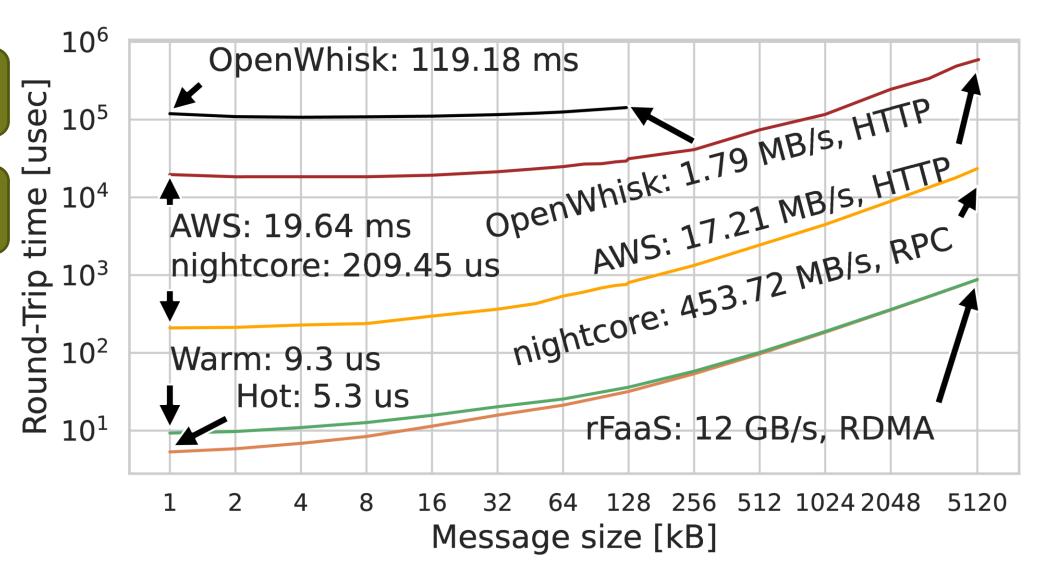






Reduced invocation critical path

Zero-copy RDMA networking











#### **FaaS in High-Performance Applications**

Serverless is slow

Answer: rFaaS

Communication is slow and restricted

Serverless is hard to program.







#### **FaaS in High-Performance Applications**

Serverless is slow



Serverless is hard to program.

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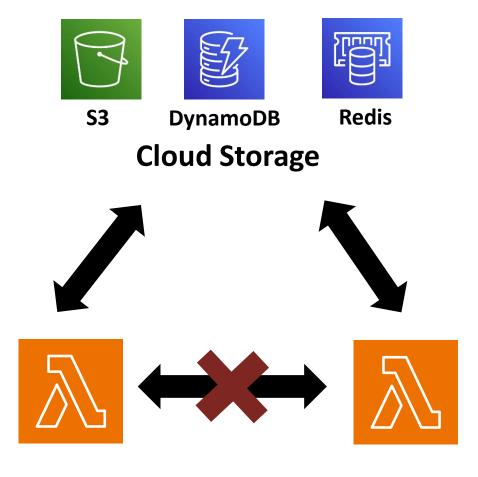






















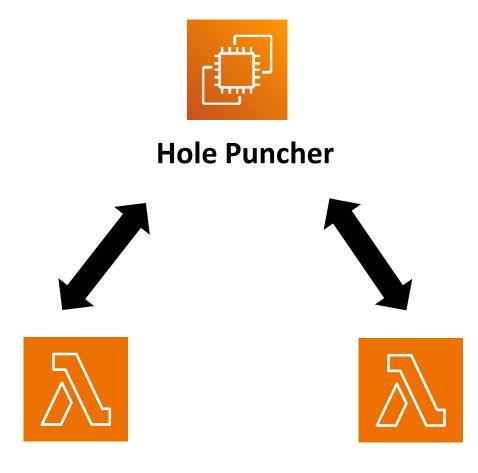










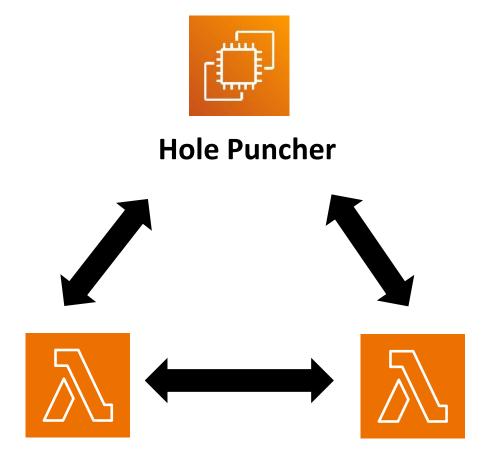












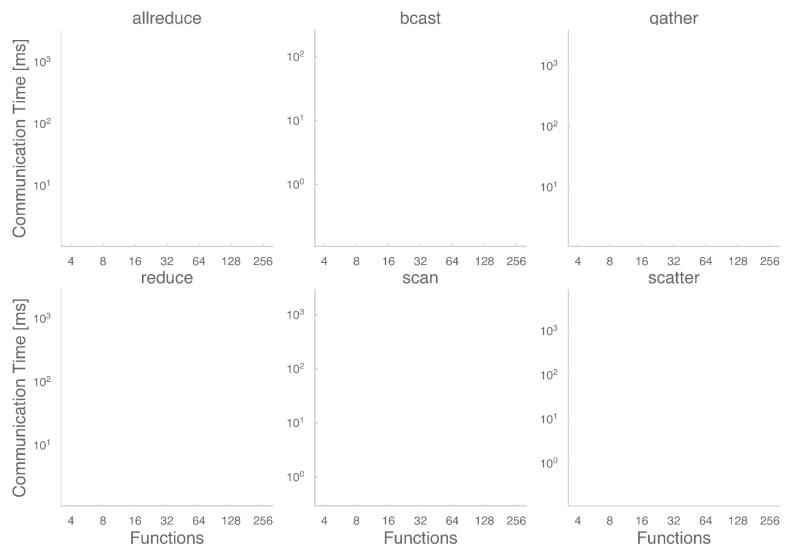








#### FMI on AWS Lambda





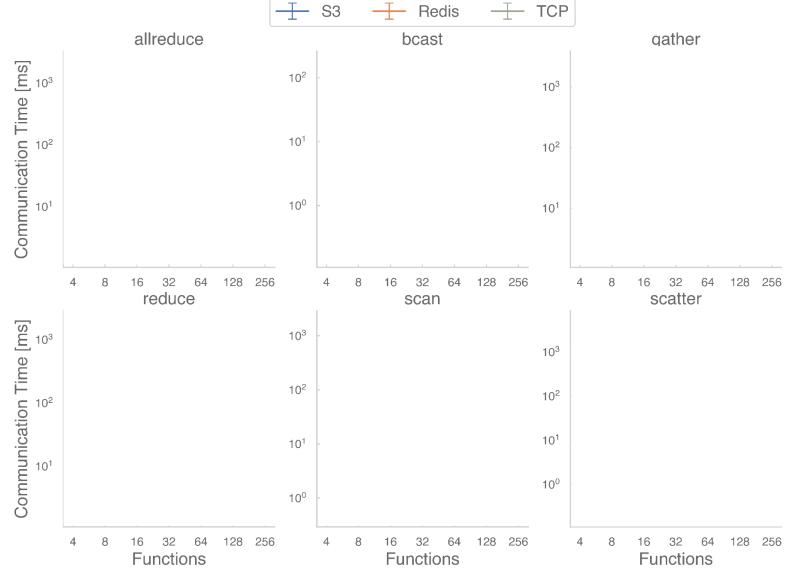
"FMI: Fast and Cheap Message Passing for Serverless Functions", ICS'23







#### FMI on AWS Lambda





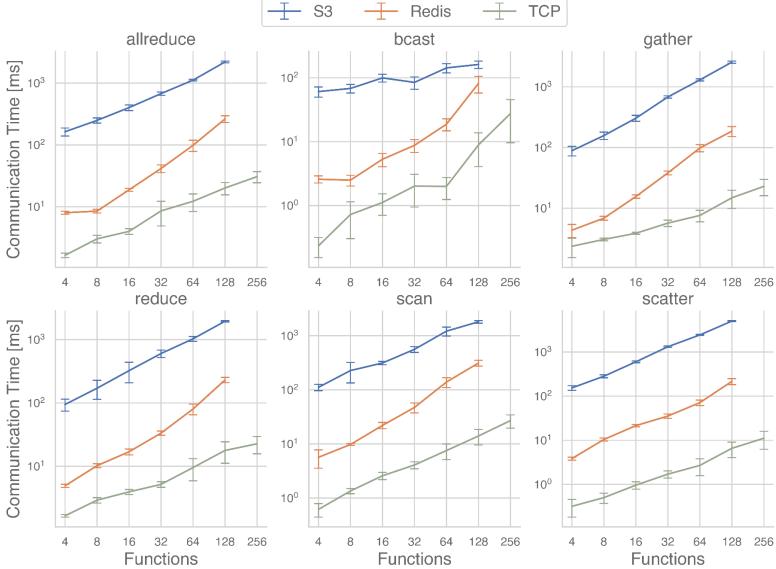
"FMI: Fast and Cheap Message Passing for Serverless Functions", ICS'23







#### FMI on AWS Lambda





"FMI: Fast and Cheap Message Passing for Serverless Functions", ICS'23







## **FaaS in High-Performance Applications**

Serverless is slow



Serverless is hard to program.

Communication is slow and restricted









## **FaaS in High-Performance Applications**

Serverless is slow

Answer: rFaaS

Serverless is hard to program.

Answer: Serverless
Processes

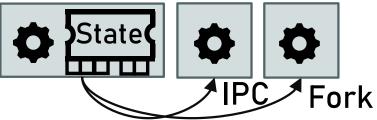
Communication is slow and restricted











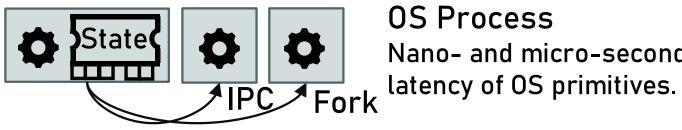
OS Process
Nano- and micro-second latency of OS primitives.



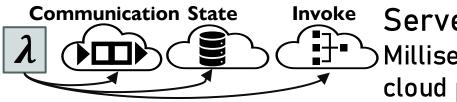








**OS Process** Nano- and micro-second



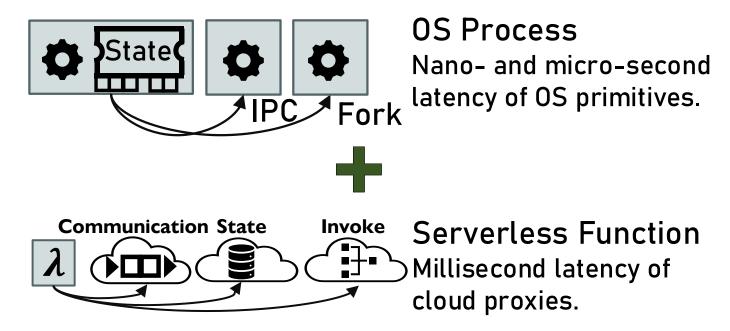
Serverless Function Millisecond latency of cloud proxies.









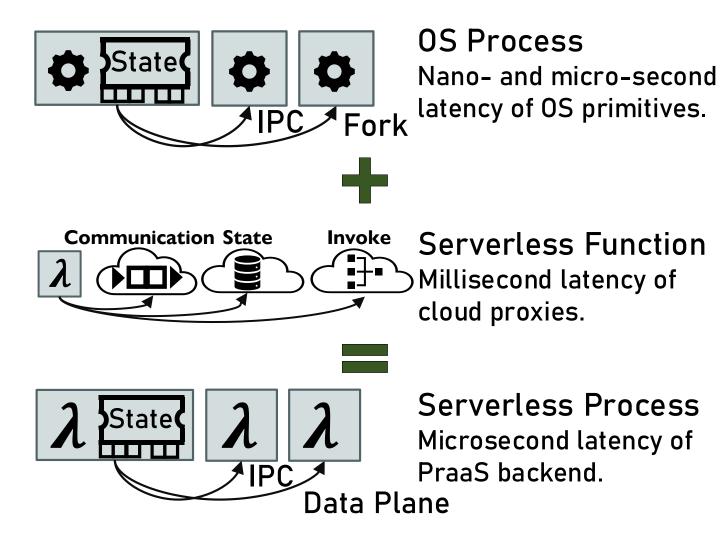










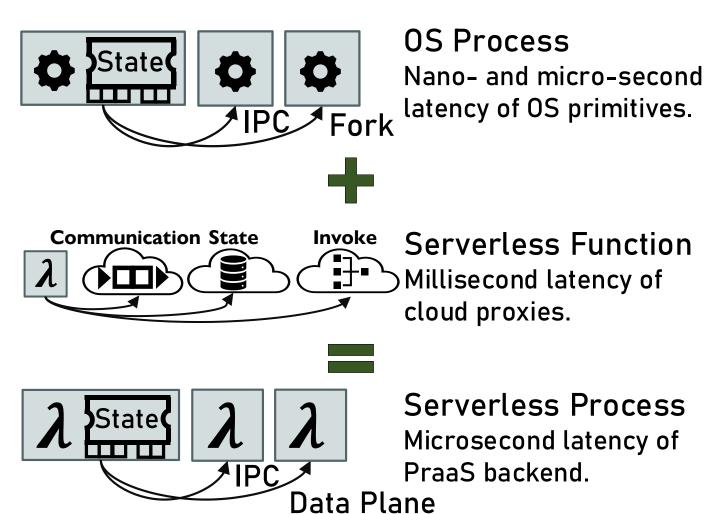














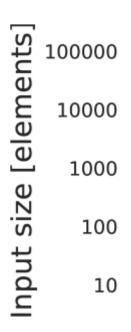
Works on AWS Fargate, Knative, Kubernetes.







#### **Reduction Benchmark: Process State vs S3**



Number of reduction invocations

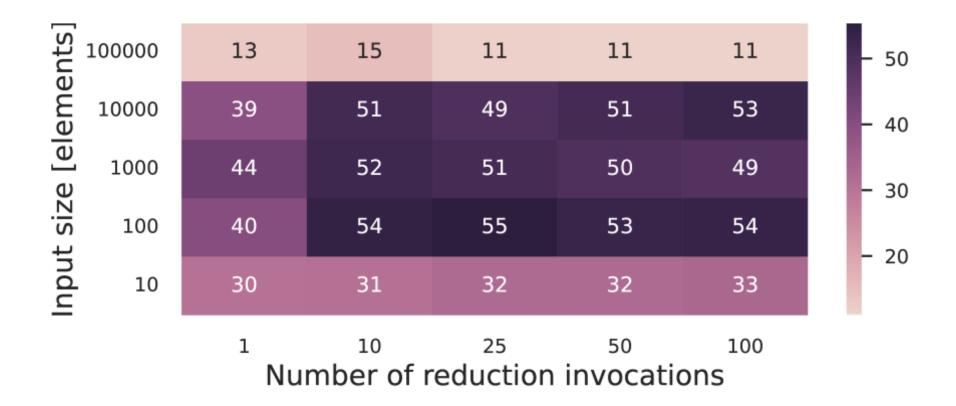








#### **Reduction Benchmark: Process State vs S3**



























( ) spcl/serverless-benchmarks











( ) spcl/serverless-benchmarks



spcl/fmi



spcl/rFaaS









( ) spcl/serverless-benchmarks



spcl/fmi



spcl/rFaaS

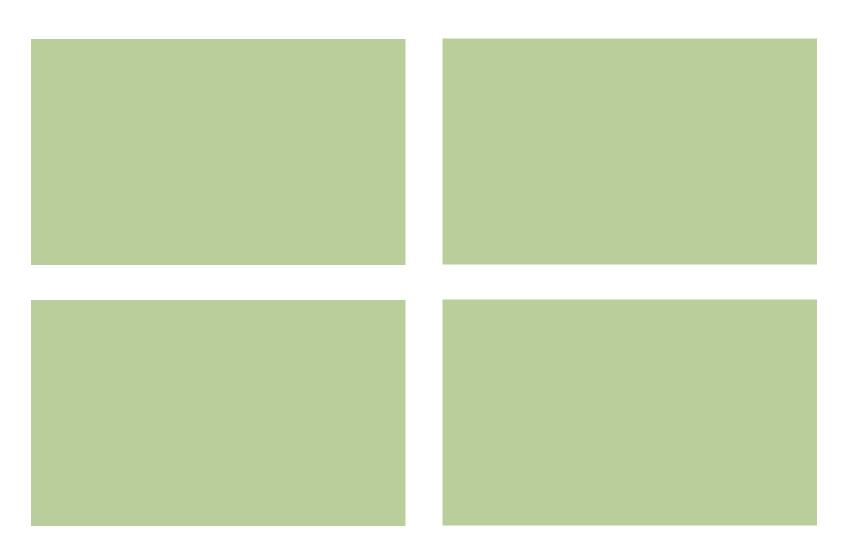


spcl/PraaS









#### More of SPCL's research:



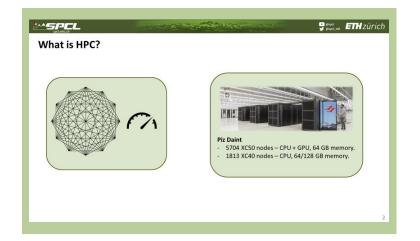
... or spcl.ethz.ch













#### More of SPCL's research:



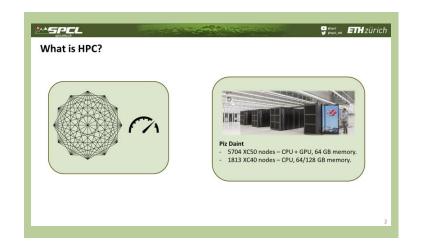
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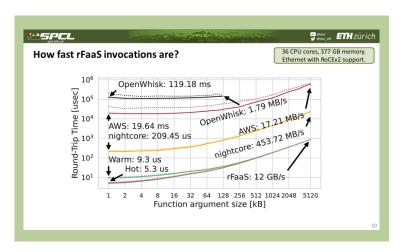












#### More of SPCL's research:



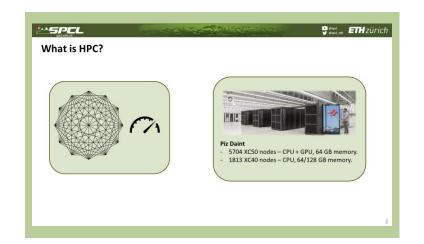
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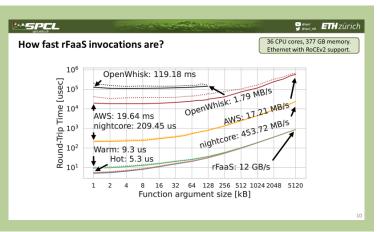


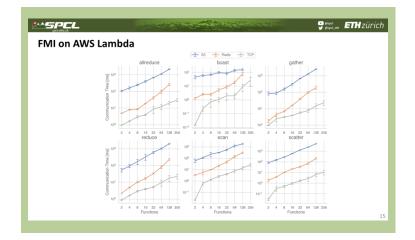


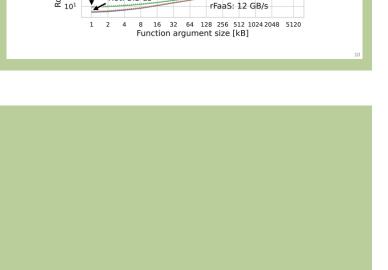












#### More of SPCL's research:



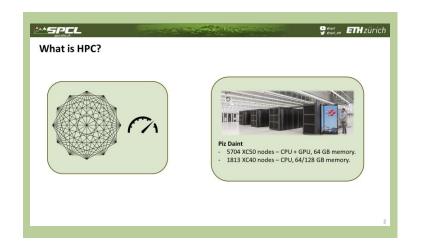
... or spcl.ethz.ch

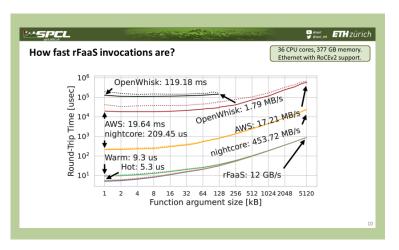




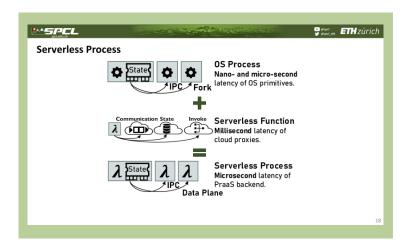








## FMI on AWS Lambda FMI on a wind to the second of the seco



#### More of SPCL's research:



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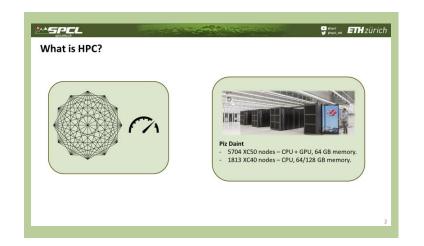
\*\*\*SPCL

FMI on AWS Lambda

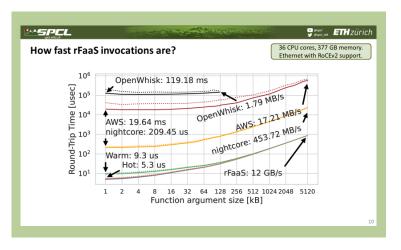




#### **Conclusions**



₽ ®sød eth **ETH** Zürich



# Serverless Process OS Process Nano- and micro-second latency of 0S primitives. Live Serverless Function Millisecond latency of cloud proxies. Serverless Process Microsecond latency of PraaS backend. Data Plane

#### More of SPCL's research:



... or <u>spcl.ethz.ch</u>



