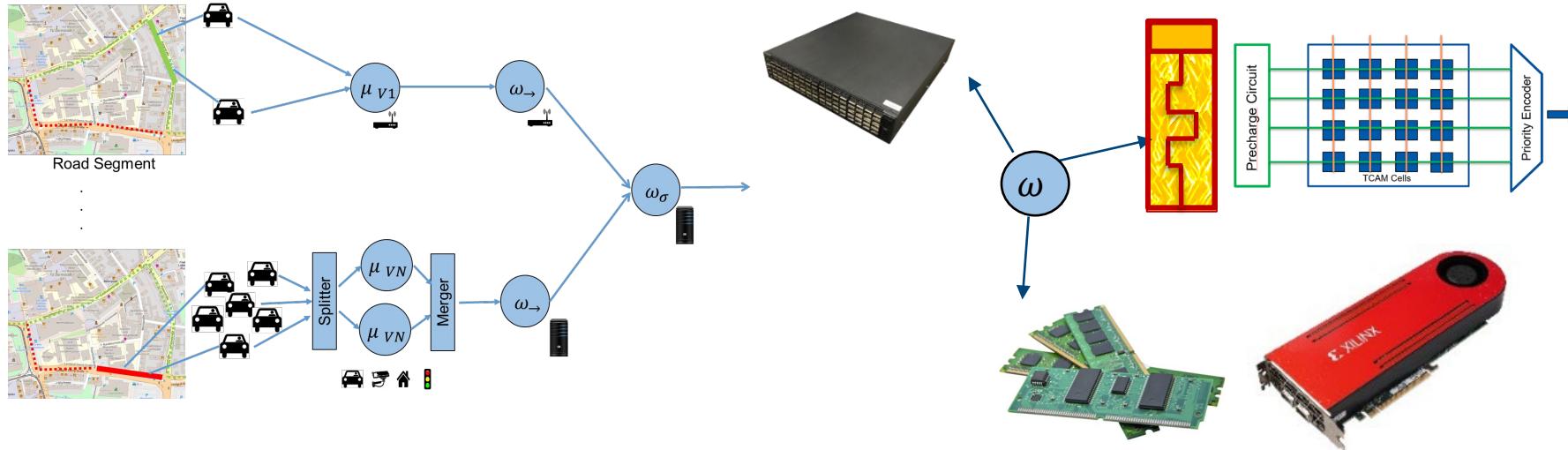


Accelerating the performance of distributed stream processing systems with in-network computing

Boris Koldehofe

@DEBS 2013



30-Jun-23

Short Introduction

Boris Koldehofe

Distributed and Operating Systems Group
Technical University of Ilmenau

Research

- Distributed data analytics
- Computer system principles
- Reliability and security

Specific Focus

- Distributed Event-based systems (DEBS)
- In-Network Computing



Data Driven Applications

Nowadays everywhere!

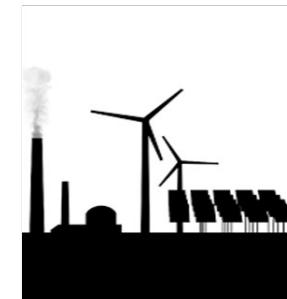
- Autonomous driving, smart factories, smart cities, telemedicine, and many more

MAPE loop of IoT services:

- Monitor and Analyze “Things”
- Plan and Execute Processes

Insights into data key to adapt applications

- Billions of things
- Exabytes of context knowledge



But Performance and Low Latency is not straight forward!

Outline

Why low latency response?

The Bottleneck in Data Movements

In-Network Computing Technologies accelerating performance

Examples in the context of Distributed Event-Based Systems

Conclusion

Low Latency responses

Often relates to highly accurate time stamps of events

Manufacturing process

- Understand correct position over time
- Low Jitter in Communications



Licensed in Adobe Stocks

Telemedicine

- Understand situations with very low reaction time



Licensed in Adobe Stocks

Financial applications

- Algorithmic trading
- Very low responses in detecting and analyzing packets
- See DEBS 2020 Grand Challenge

Improving Timestamp Accuracy Technological developments

5G and even 6G Campus networks

- Goal interconnect processors fast
- 100 μ s - 1ms delays, high mobility

TSN

- Real-time guarantees for industrial applications

Edge Computing

- Offload Computations

Accelerators

- Computation
- I/O
- Protocols / Architectures

Timestamp inaccuracy	Location Inaccuracy
1s	10m
1ms	10cm
1 μ s	0.1mm

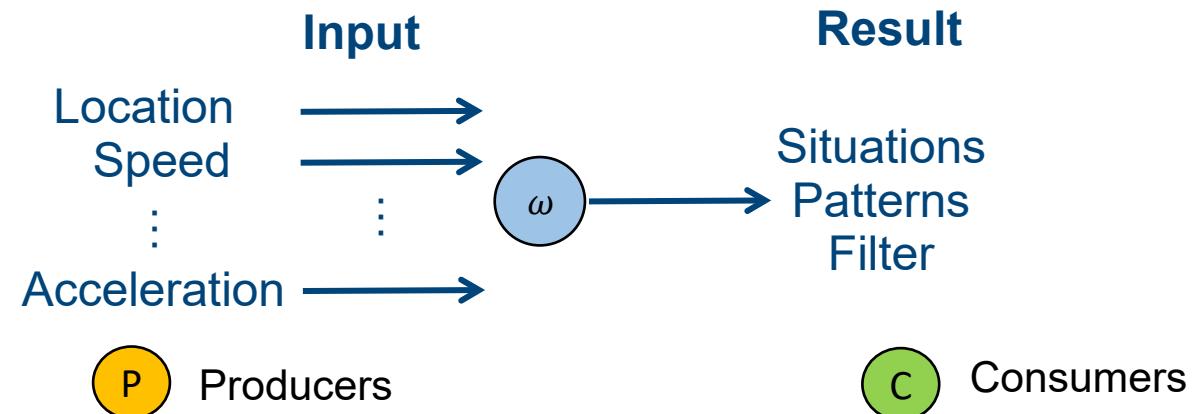
Moving Object of 36km/h



DEBS / Real-Time Analytics

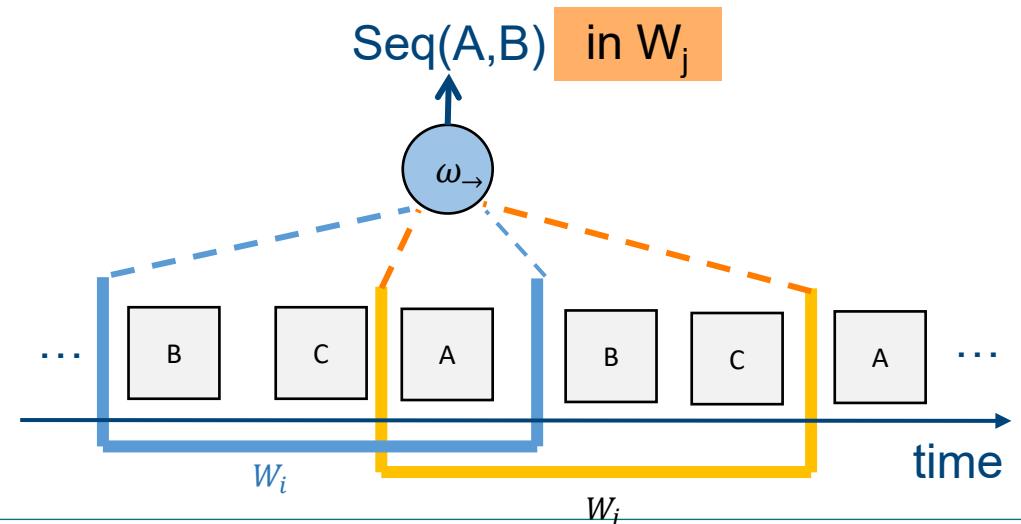
Correlations on data stream

- With low end to end delay
- High accuracy detection



Paradigm:

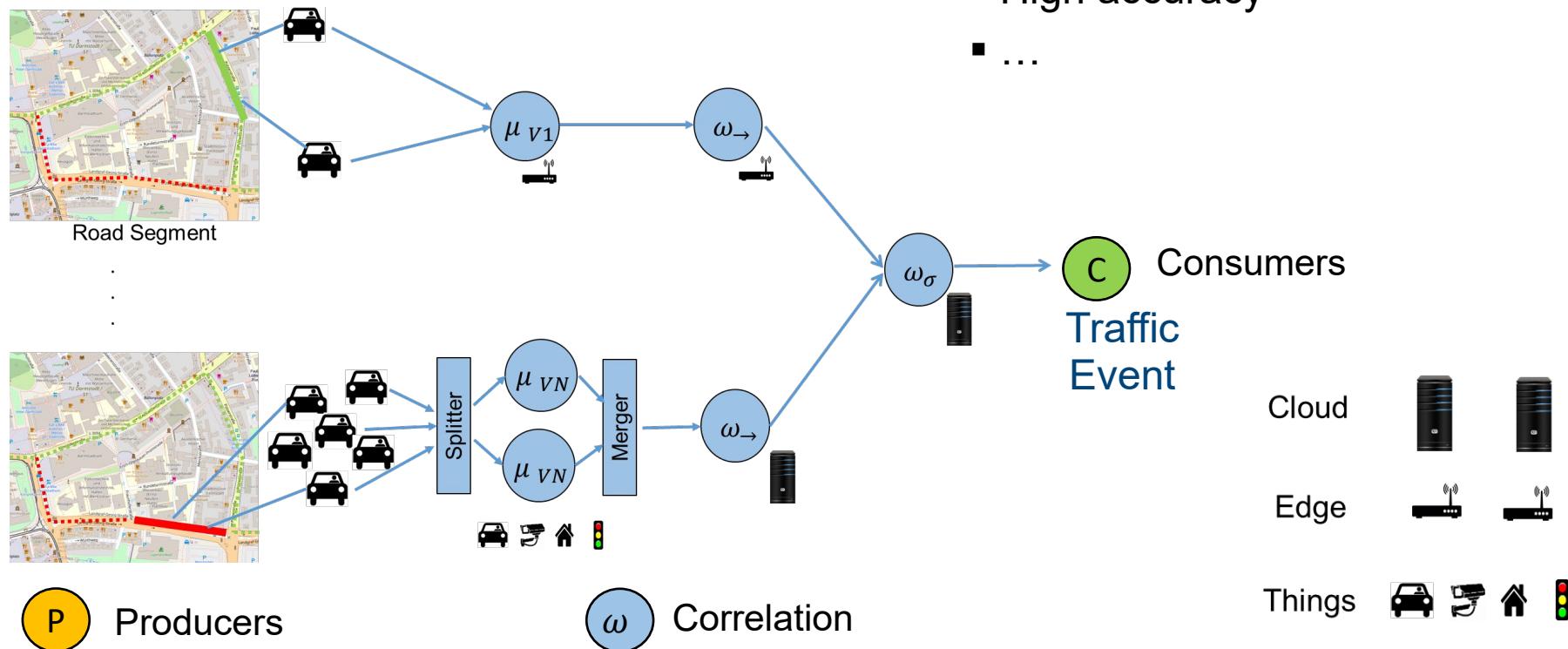
- Operators identify pattern on partial data stream: *window*
- E.g. CEP operator, Filter, Neural Network, Deep Learning Model



Distributed Real-Time Analytics

Execute operator network on a distributed infrastructure

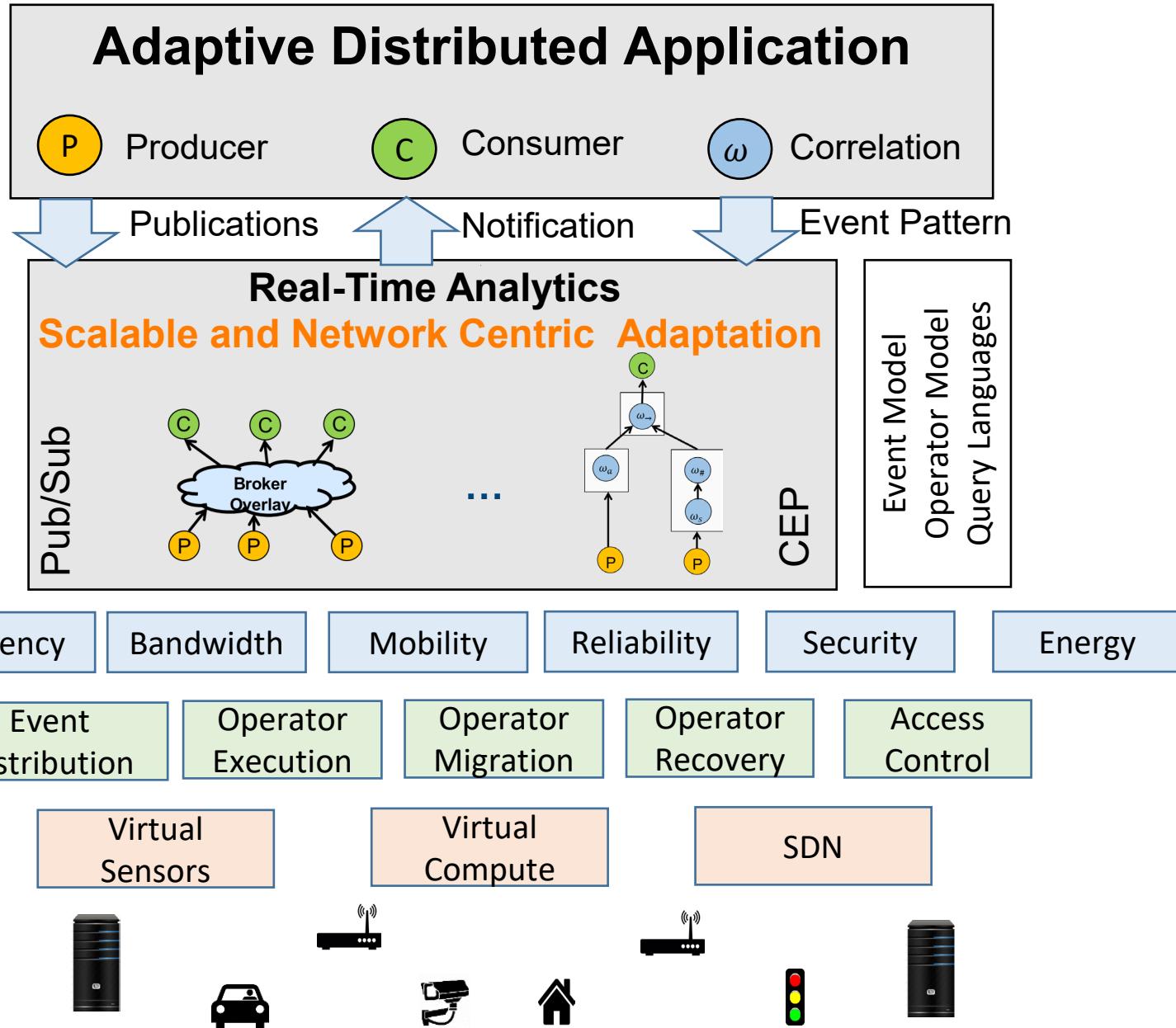
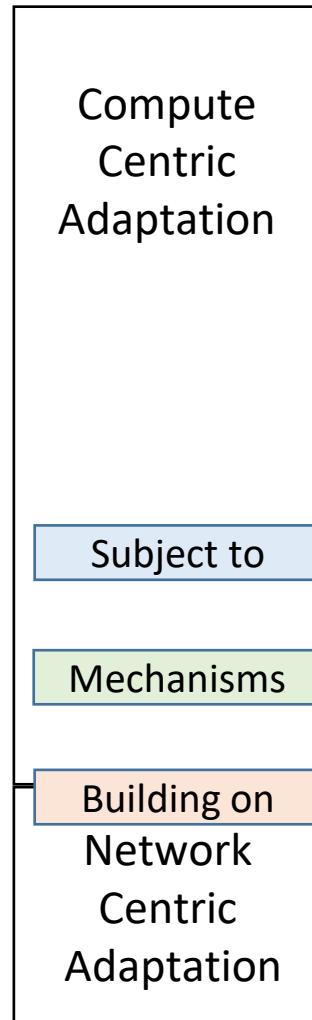
Increase Scalability and Performance



Optimization subject to potentially conflicting goals

- Decoupling producers and consumers
- Low end-to-end delay
- High accuracy
- ...

Adaptive Distributed Application

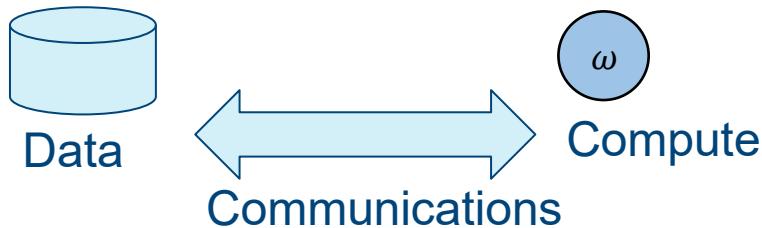


Meeting Performance of Time Sensitive Distributed Applications

Cyberphysical application

- Low latency?
- Predictable performance?

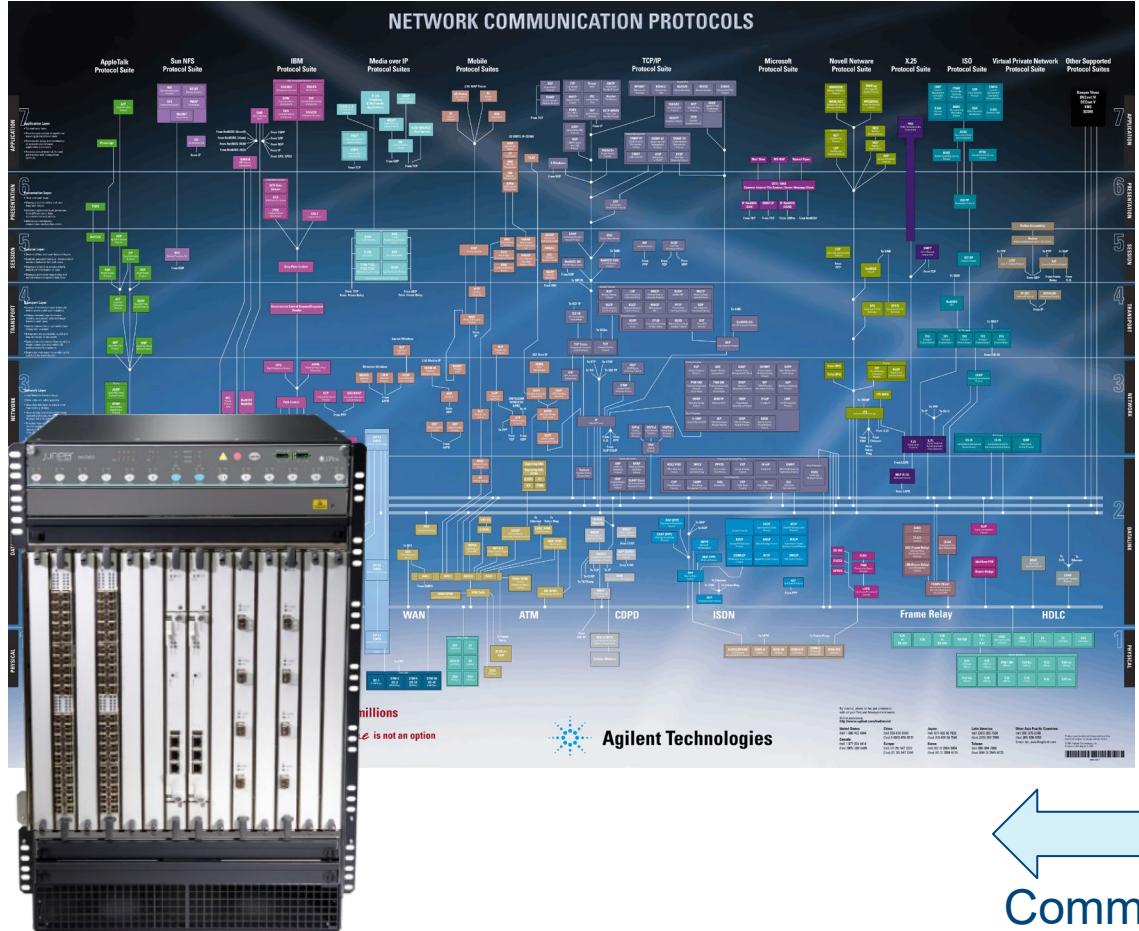
Bottlenecks in data movement and processing



Requires much more flexibility in using mechanisms of the distributed infrastructure!

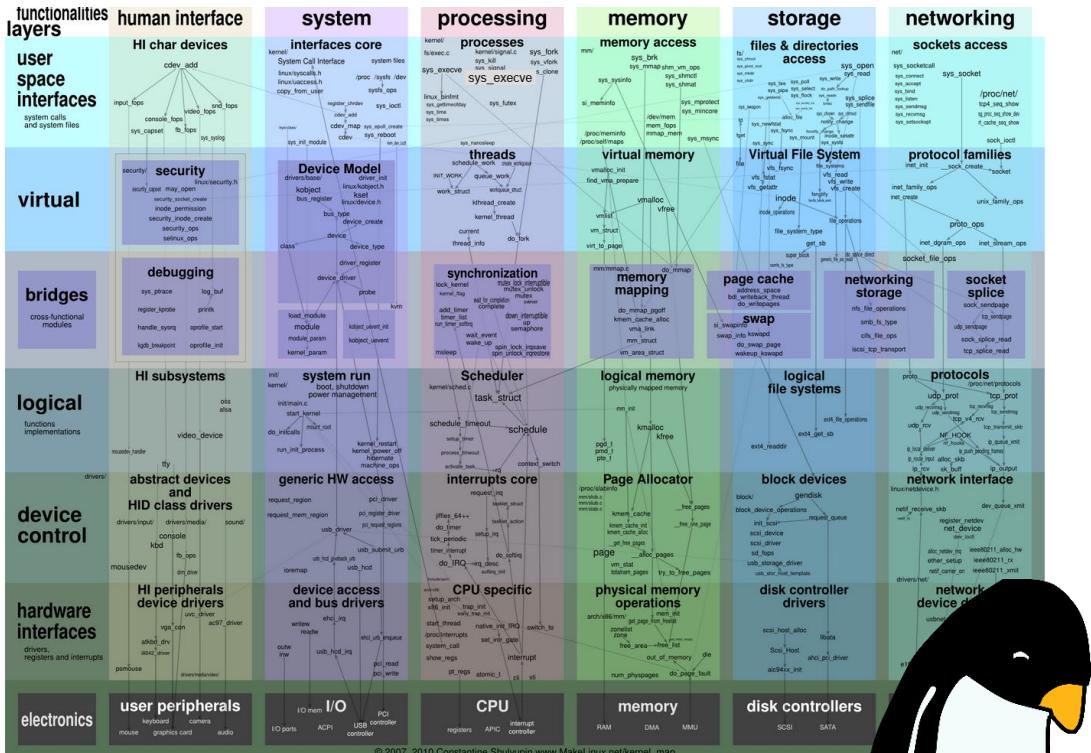
Lack for Flexibility: Communication Protocols and Operating Systems

Hardcoded in network appliances



Communications

Time for data to bypass the kernel



Ingredients for Increased Flexibility

Programable hardware

- P4 Switches
- NetFPGA

New networking paradigms

- Software-Defined Networking
- Network Function Virtualization

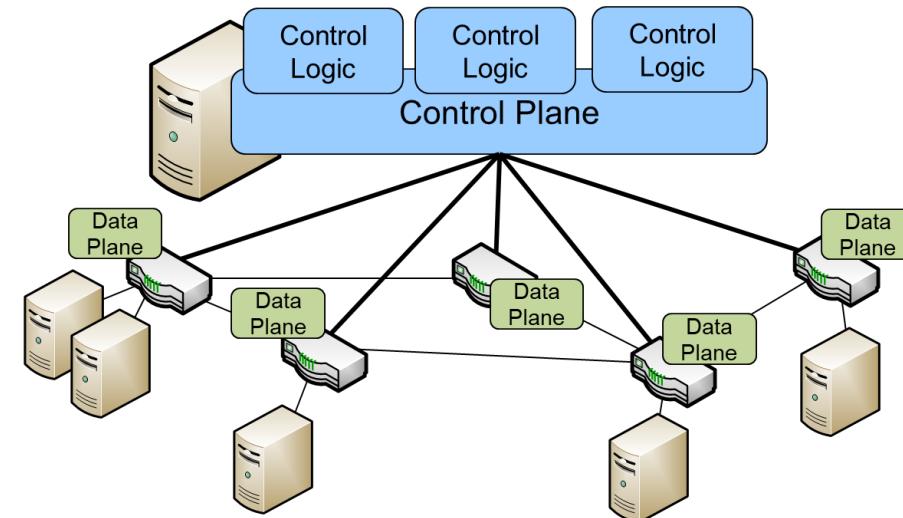
Significant changes in the infrastructure

- Edge Data Center
- Technologies & Concepts
- DPDK, P4, OpenFlow, RDMA



Barefoot Tofino

FPGA



Enabler for in-network computing!

In-Network Computing

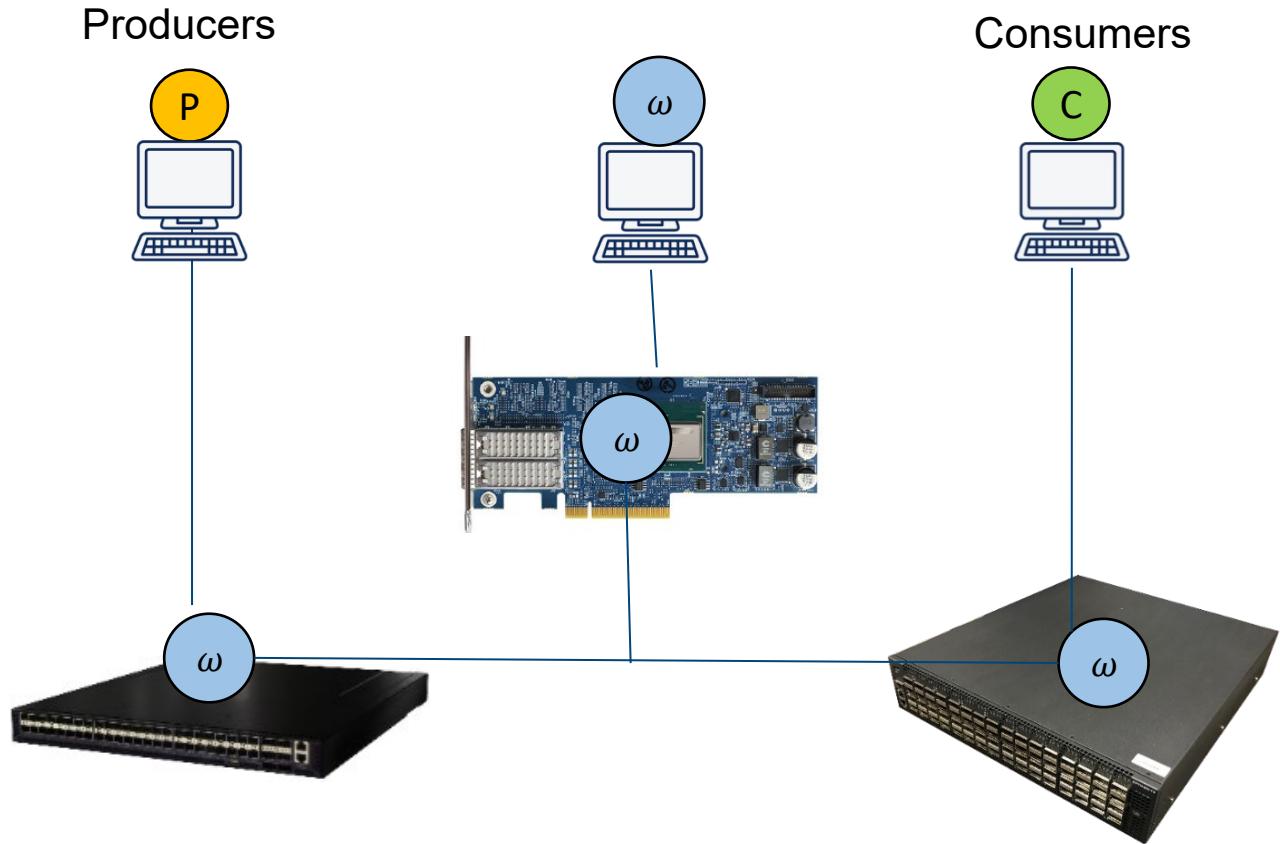
Idea enable computations on the data path

Traditionally,

- Packet header processing,
 - e.g., routing, firewall, packet classification, load balancing, deep packet inspection

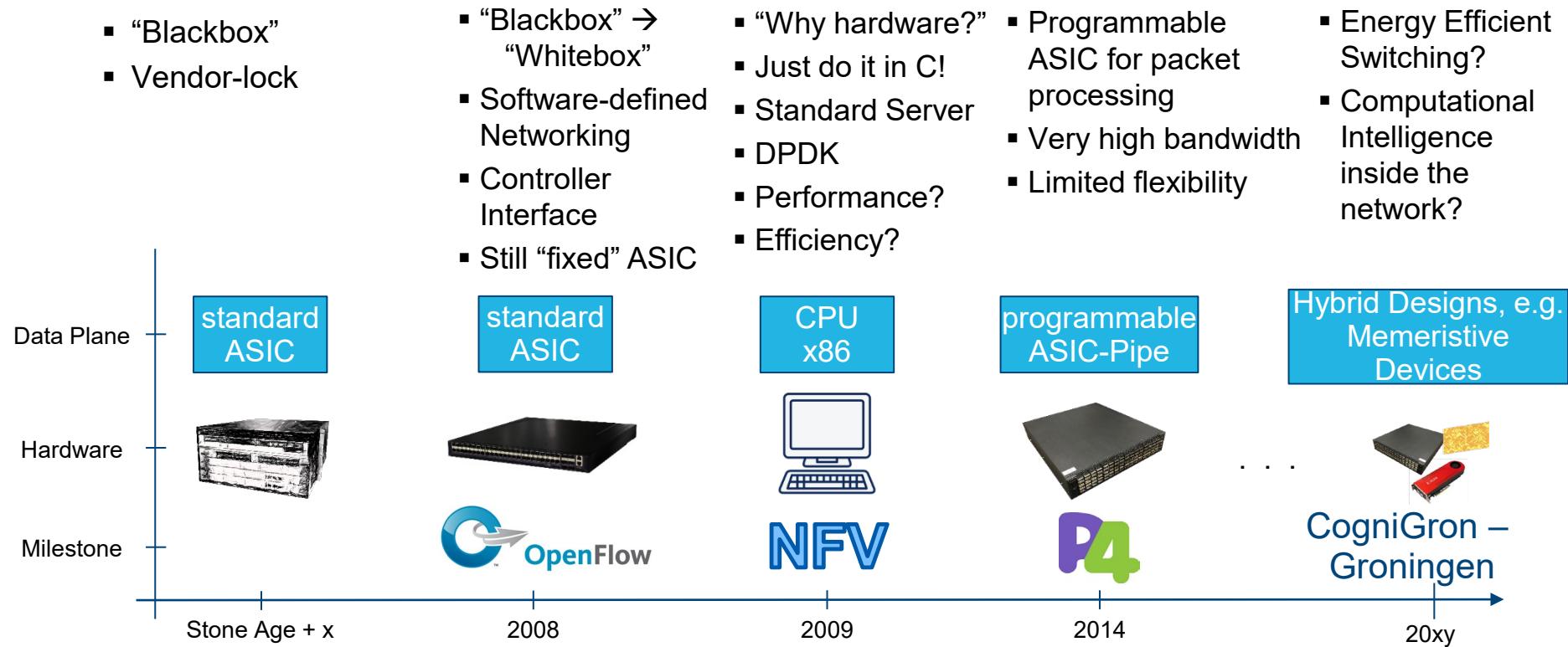
Often

- Match/action pipeline model of networking hardware
- Management interface, specific programming interfaces, ...



Evolution: In-Network Computing resources

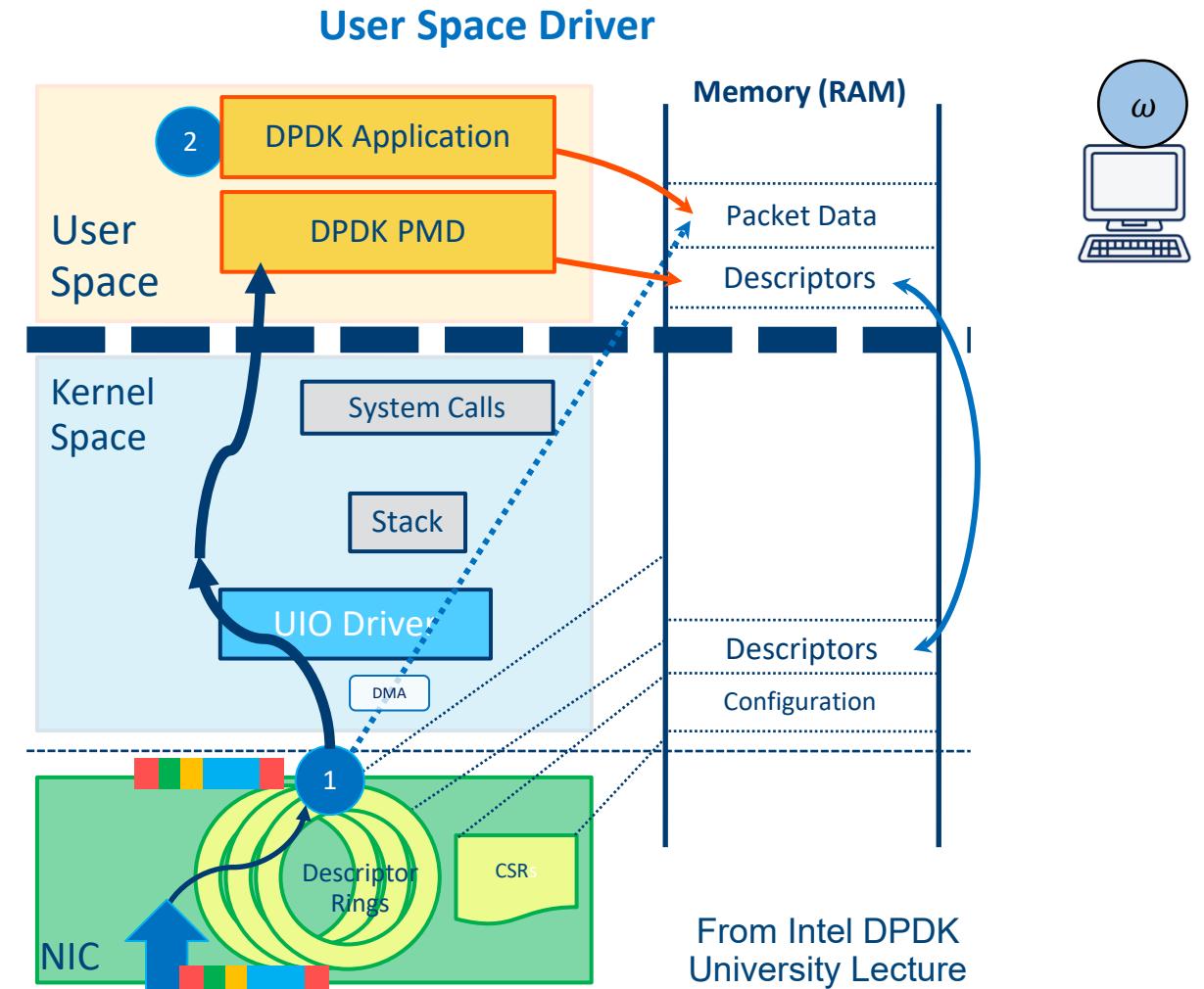
Towards flexible, high performance, and energy-efficient in-network computing



Performance Acceleration via INP

INP resources can reduce the time to move data, e.g.

- DPDK: circumvent OS
- OS Kernel: Enhance Communication Protocol
- NIC: process ahead of OS
- Switch : closer to producer/ consumer



Performance Acceleration via INP



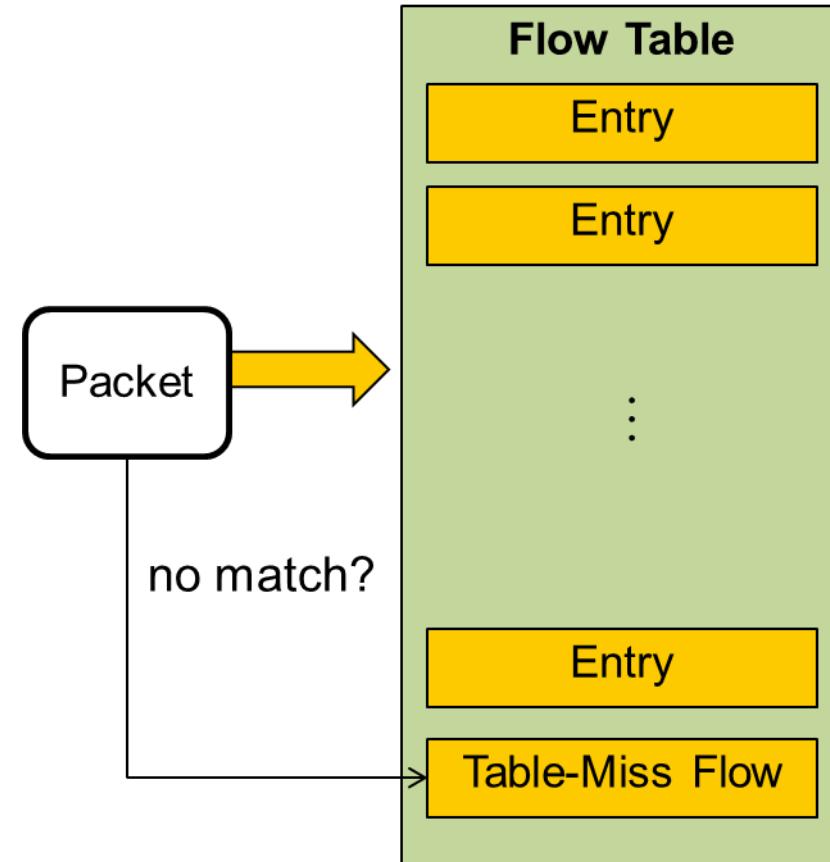
INP resources can reduce the time to move data, e.g.

- DPDK: circumvent OS
- OS Kernel: Enhance Communication Protocol
- NIC: process ahead of OS
- Switch : closer to producer/ consumer

INP resources can accelerate the processing time

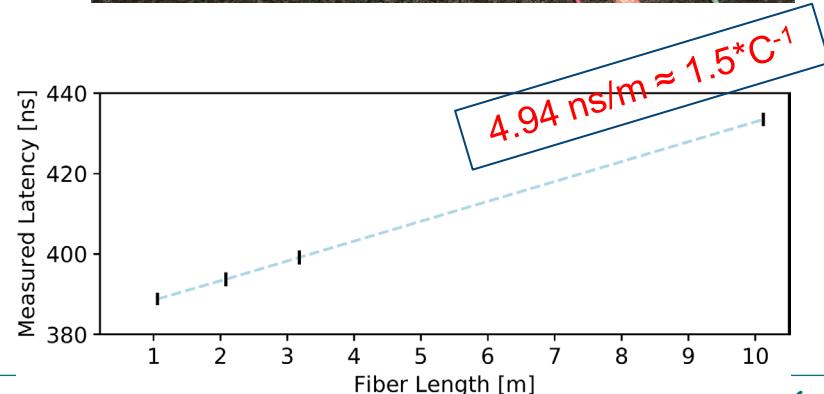
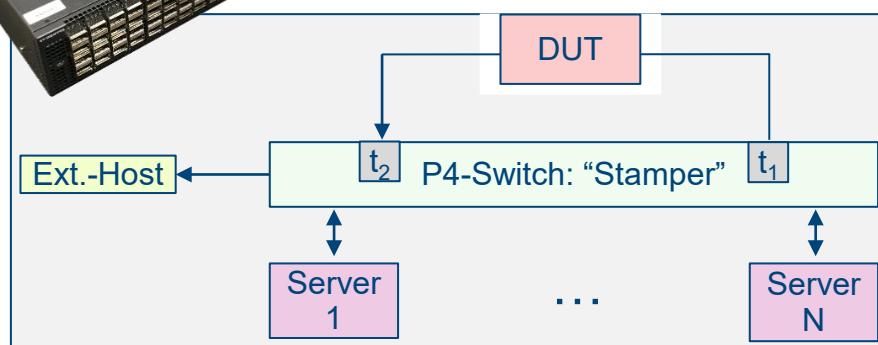
- Efficient Matching : TCAM
- Transformation and routing

INP enables dynamic exchange of functionality



Is INP = Low Latency? High Performance Packet Analytics in P4STA

	actual fiber length	avg. latency	std. dev.	loss
1 m	1.06 m	107.830 ns	1.46 ns	0 packets
2 m	2.08 m	112.850 ns	1.61 ns	0 packets
3 m	3.18 m	118.336 ns	1.52 ns	0 packets
10 m	10.12 m	152.883 ns	1.61 ns	0 packets



Challenges in using them for Real-time analytics

Specific domain specific programming models

- OpenFlow, P4, Verilog

Breaking distribution transparency

- E.g., applications does not work on byte streams, but packets!

Increased heterogeneity

Headers may leak information on the packet content

Outline

Why low latency response?

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Examples in the context of Distributed Event-Based Systems

Conclusion

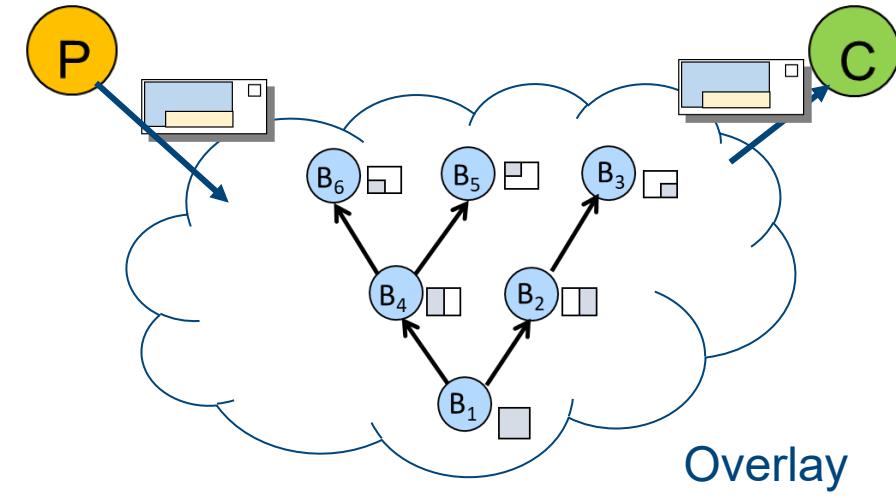
Publish/Subscribe and Performance

Efficient distribution by means of overlays

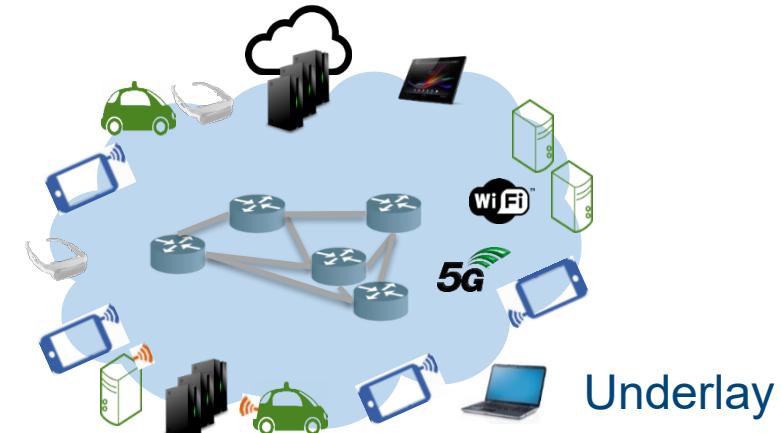
Bandwidth efficient overlays

BUT big performance gap

- Overlay
- Underlay



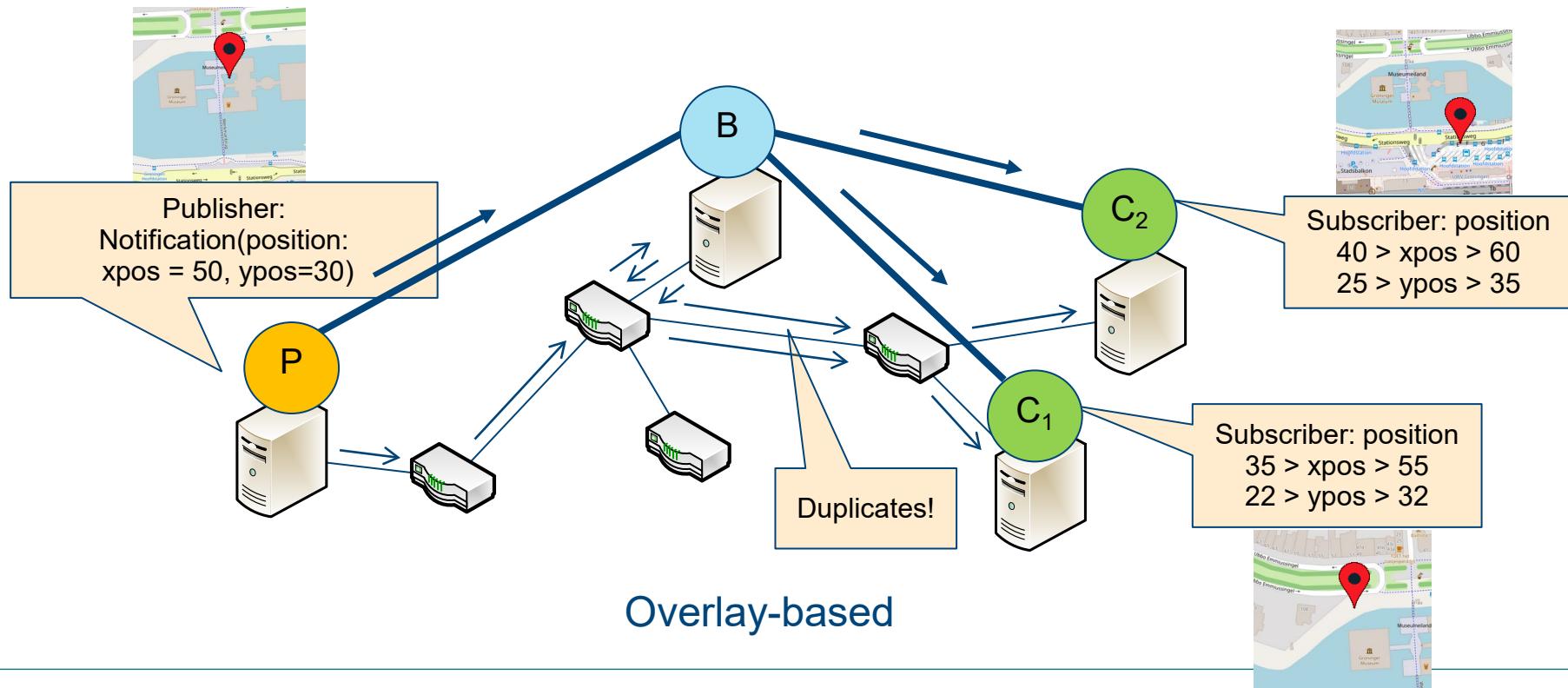
vs



High Performance Publish/Subscribe: Basic Idea

Reduce the overhead:

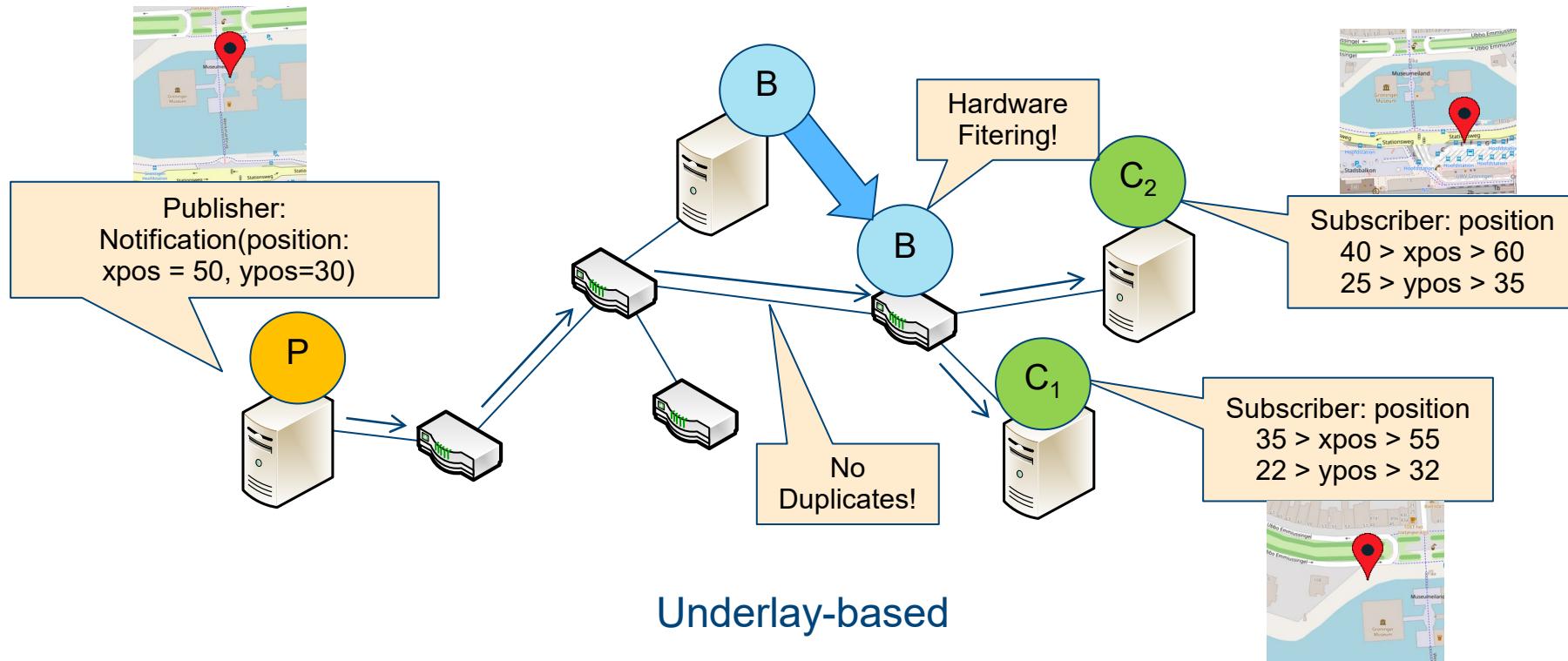
- Message duplications
- Matching subscriptions at the hardware



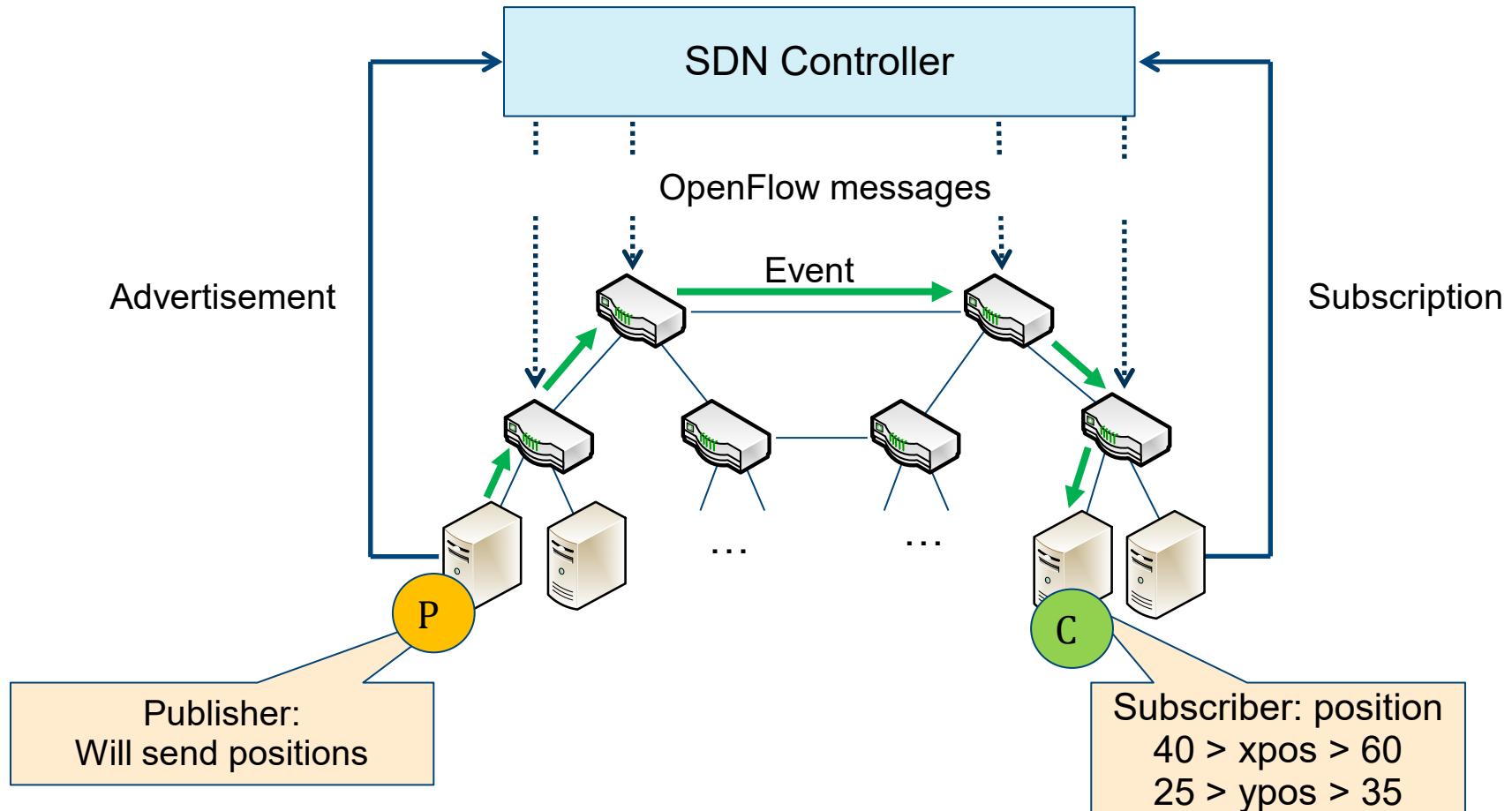
High Performance Publish/Subscribe: Basic Idea

Reduce the overhead:

- Message duplications
- Matching subscriptions at the hardware



SDN-based Publish/Subscribe Middleware

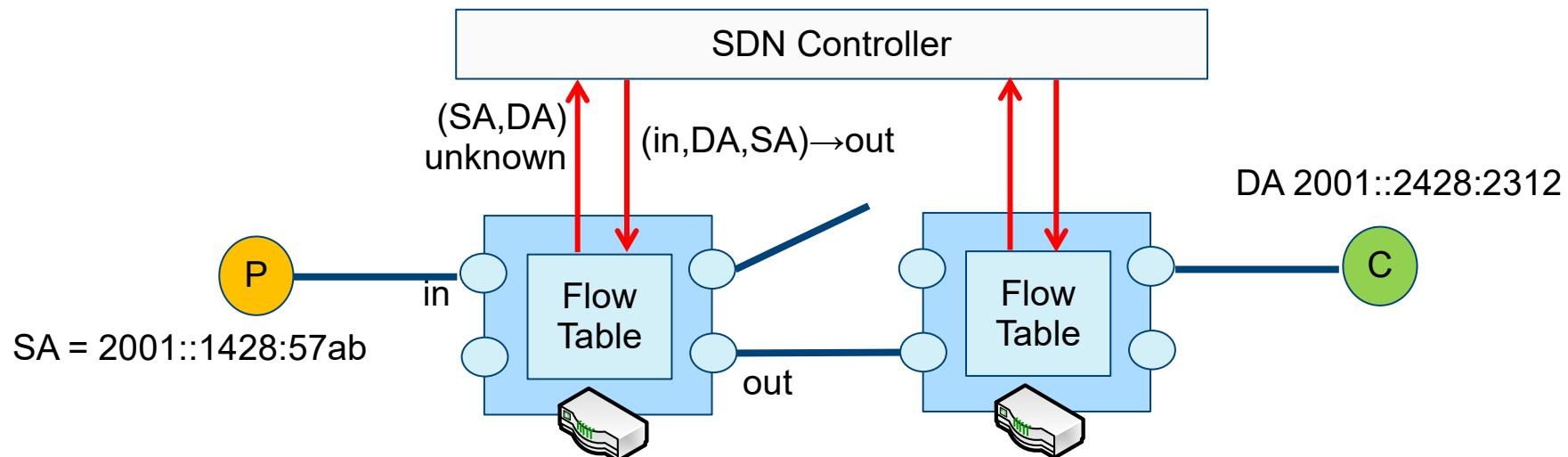


Configuration Based on OpenFlow

Forwards packets from *in ports* to *out ports* by means of flow table, e.g.,

In port	VLAN ID	Ethernet			IP			TCP	
		SA	DA	Type	SA	DA	Prot	Src	Dst

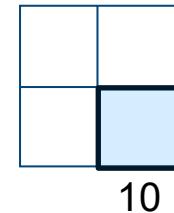
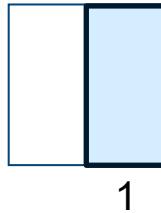
Controller can add, change and remove flow entries using OpenFlow



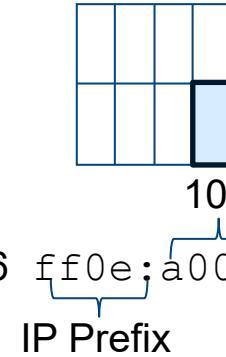
RQ: How to represent and match content-based subscriptions, e.g. in OpenFlow?

Subscription and event matching in flow table

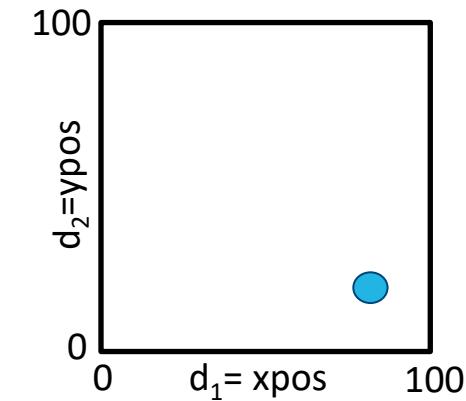
1. Generate binary representation based on spatial indexing
2. Map binary representation to IPv6 Multicast address
 - Coexistence with other services



Mapping to IPv6



IP Prefix



Approach overview

Subscription/Advertisement

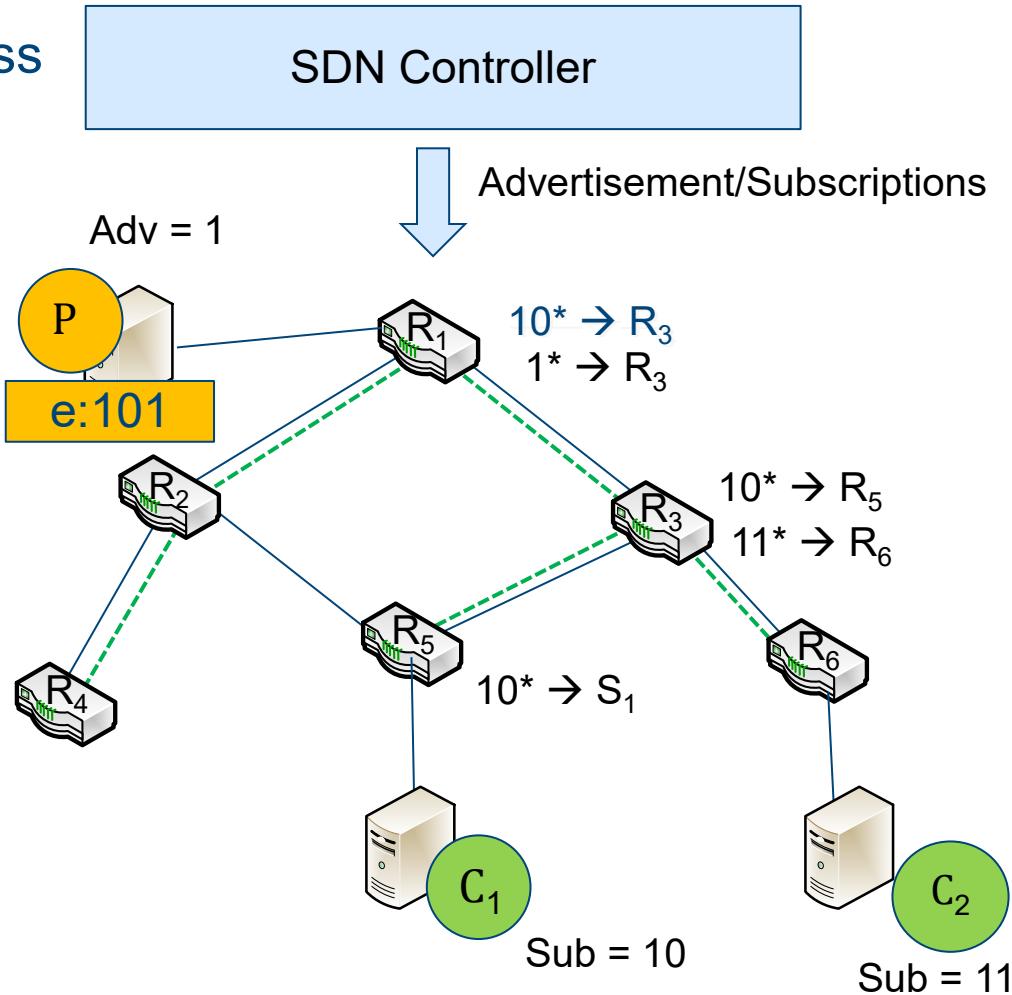
- Sent to controller with predefined IP address

Controller optimizes topology

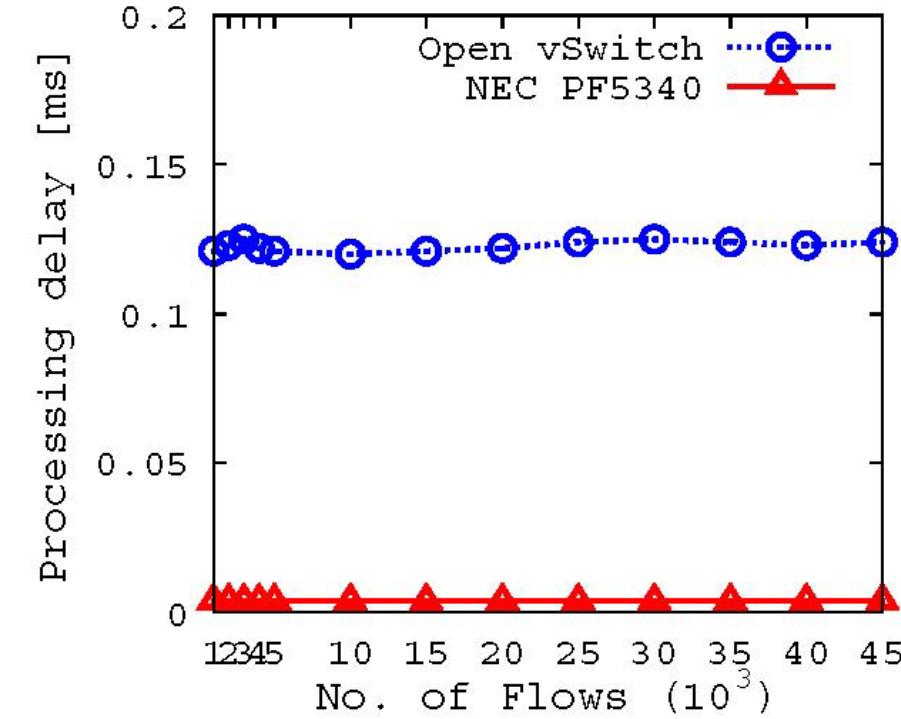
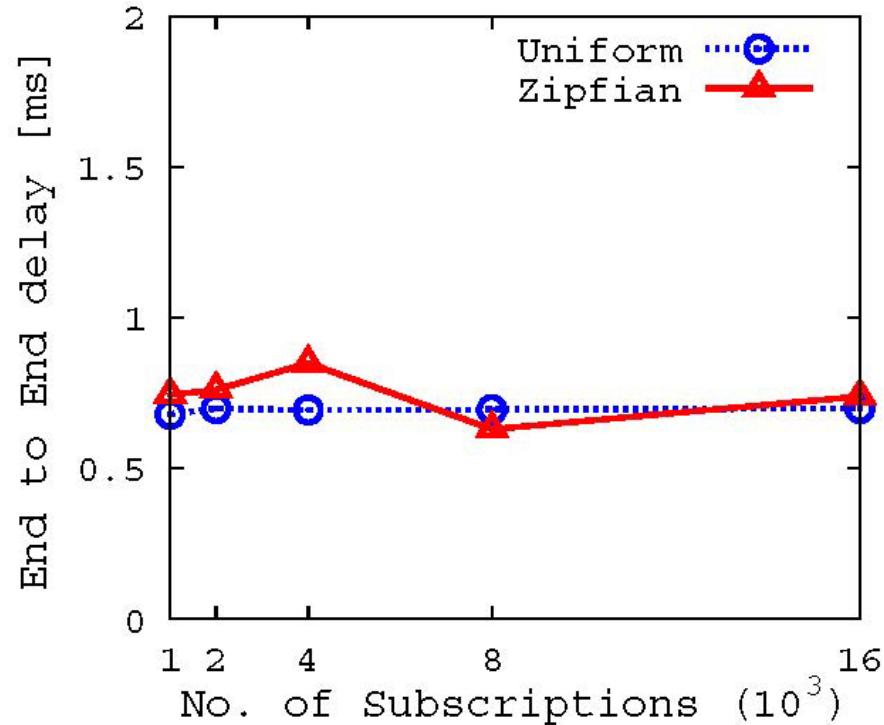
- Establish paths between publishers and subscribers
- Paths are established along a tree

Events

- Directly sent to the network
- $IP_{Prefix} \circ$ bit string



Result: Forwarding performance



Hierarchical fat-tree topology
10 Open vSwitches and 8 end-hosts
10,000 events

Properties

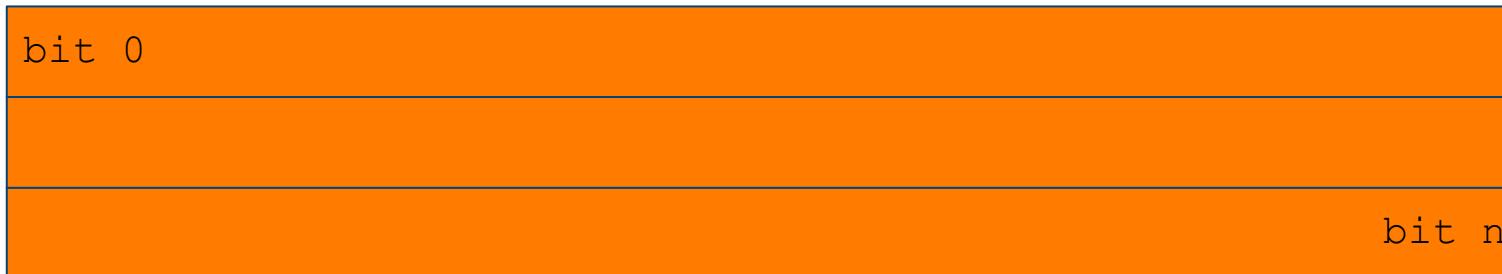
OpenFlow-based Management enables expressive subscription management

But requires from every publisher/ subscriber

- Understand the encoding

Relied on specific Header Fields!

But would work in general using a big field or mask



More Complex, state-full not considered!

Extending to P4 based INP



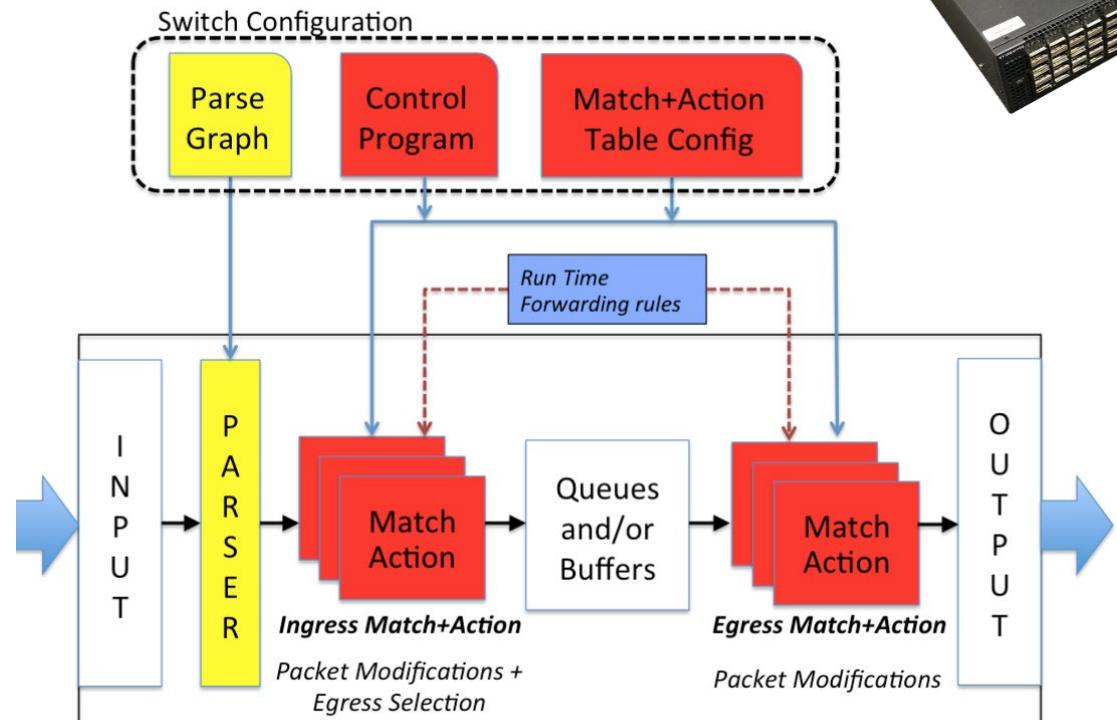
P4 supports Programming Reconfigurable Match Action Pipeline

Define own Protocol Headers to be used by DEBS

Define Matching Operations for specific Header fields

```
typedef bit<32> timestamp_t;
typedef bit<16> type_t;
typedef bit<8> attribute_t;
typedef bit<8> value_t;
```

```
header event_h {
    type_t type; /* example: weather. */
    timestamp_t timestamp; /* event occurrence time. */
    attribute_t attribute; /* example: humidity, temperature. */
    value_t value; /* example: 45% humidity and 23 degrees celsius temperature.
}
```



[Source p4.org](http://p4.org)

P4: Enhancing Stateful Operations

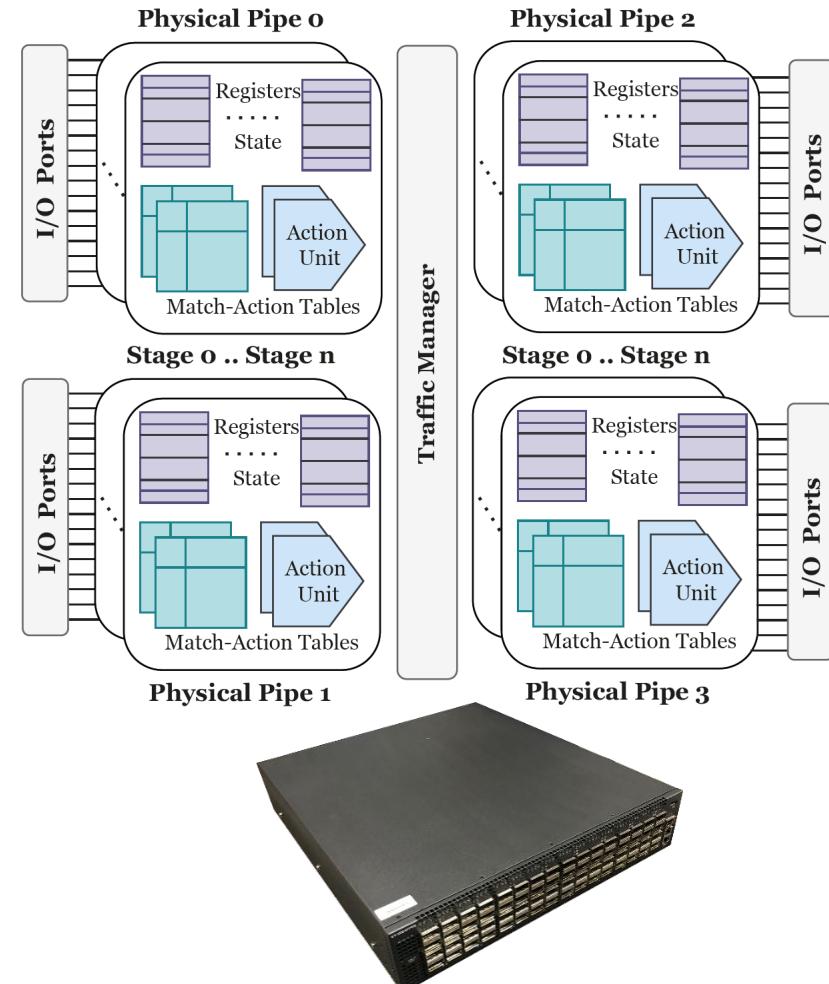
Limited Support for Stateful Operations

Many pitfalls:

- No sharing of registers between different stages of the pipeline
- Exclusive read or write operations
- Packet cannot iterate over all registers

However, can be used to model for specific platforms stateful CEP operators!

Kohler, Mayer, Dürr, Maaß, Bhowmik, and Rothermel. *P4CEP: Towards In-Network Complex Event Processing*. In Proceedings of the 2018 Morning Workshop on In-Network Computing (NetCompute '18, pp. 33–38).
<https://doi.org/10.1145/3229591.3229593>

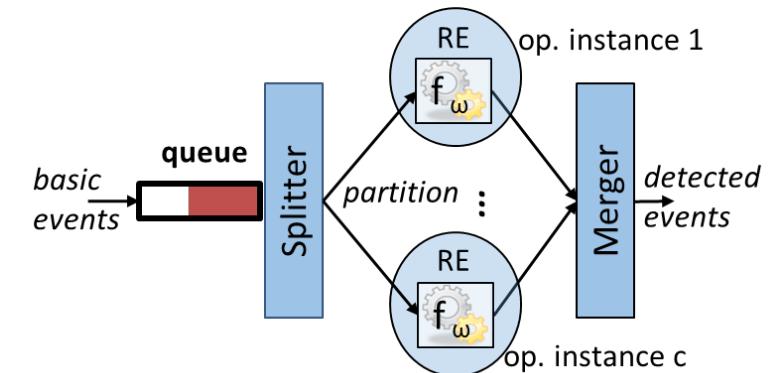


Supporting parallel operator execution with P4

Operator Parallelization is a common method in DEBS

Splitter:

- Partition streams in independent processable windows
- Operator instances return results to the merger
- Merger coordinates streams



Processing rate of the splitter is the bottleneck in scaling operators

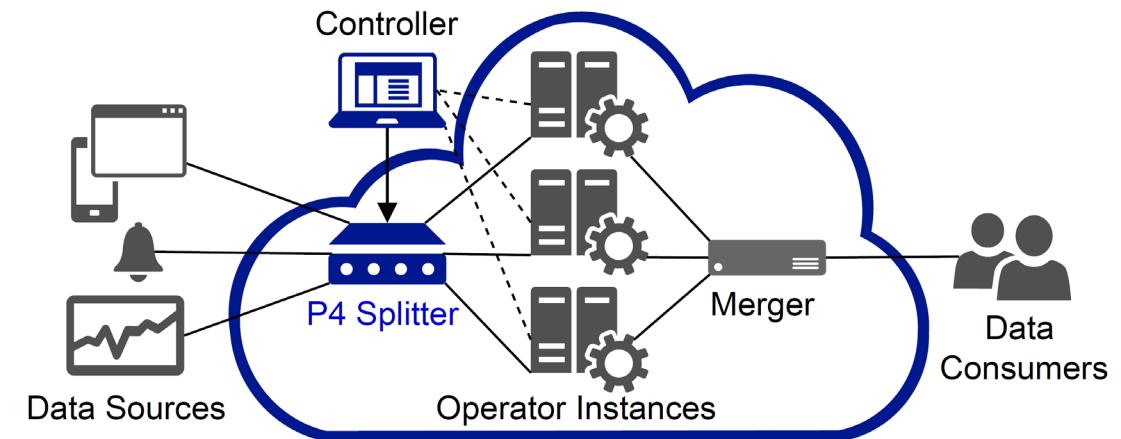
Can be done already on the path between producers and consumers

P4 Splitter: Window Operators

Idea: perform stream partitioning via INP

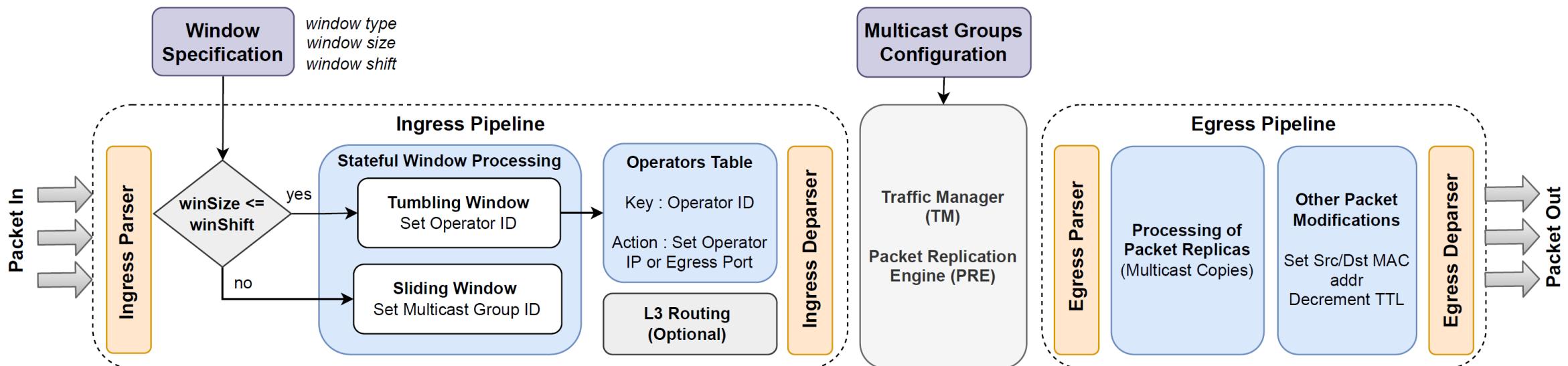
Problems:

- Dynamically expressing multiple distinct window semantics for operators
 - Time-based, Count based, ...
- Needs to be performed in line-rate with
 - Match Action Logic
 - Registers state



Basic Idea / Procedure

1. Each stream identified by an id → Matching events
2. Window specifications can be dynamically added/removed
 - with respect to a unique stream id (Dynamically Matching rules)
3. Window state is captured via registers
4. Incoming events trigger updates to window state (dependent on window)
5. Will be added to a multicast group that sends an event packet to all destinations

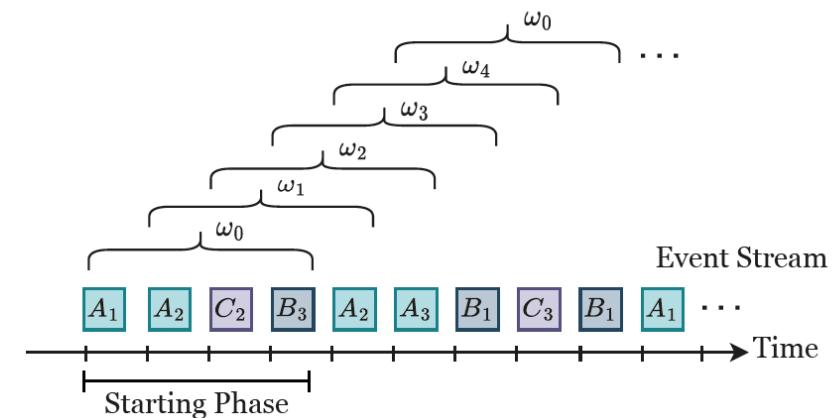


Example

Specifying a window semantics:

- e.g. Count-based Sliding Window
 - stream ID idx
 - parallelism degree N
 - window size n
 - window shift δ , $\delta \leq n$ (counting)

idx	latest Op.	$Overlap$	Op. Inst
0x23	ω_0	0	ω_0
0x23	ω_1	1	ω_0, ω_1
0x23	ω_2	2	$\omega_0, \omega_1, \omega_2$
0x23	ω_3	3	$\omega_0, \omega_1, \omega_2, \omega_3$
0x23	ω_4	3	$\omega_1, \omega_2, \omega_3, \omega_4$
0x23	ω_0	3	$\omega_2, \omega_3, \omega_4, \omega_0$
0x23	ω_1	3	$\omega_3, \omega_4, \omega_0, \omega_1$
0x23	ω_2	3	$\omega_4, \omega_0, \omega_1, \omega_2$



Round-robin load-balancing over 5 operator instances with $n = 4$, $\delta = 1$.

Challenge: How to evaluate such a system?

Although P4 facilitates programming INP hardware, it is time consuming

One device costs ~5000-10000€

Huge Development effort

Don't expect large scale comparison or a baseline comparison with Apache Flink

In general baseline comparisons

- Being faster neither straight forward nor very insightful

What did we evaluate:

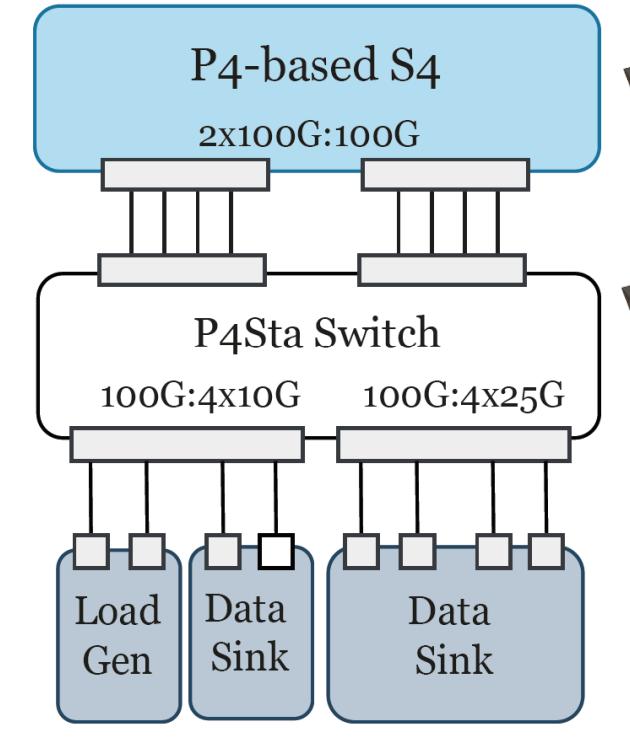
P4STA for Packet generation and validation

What is the latency introduced by a INP

Measure feasible throughput

Measure resources

- How many streams, operators, and windows can be supported



Some Findings

Throughput and Low Latency

High Throughput low and stable latency

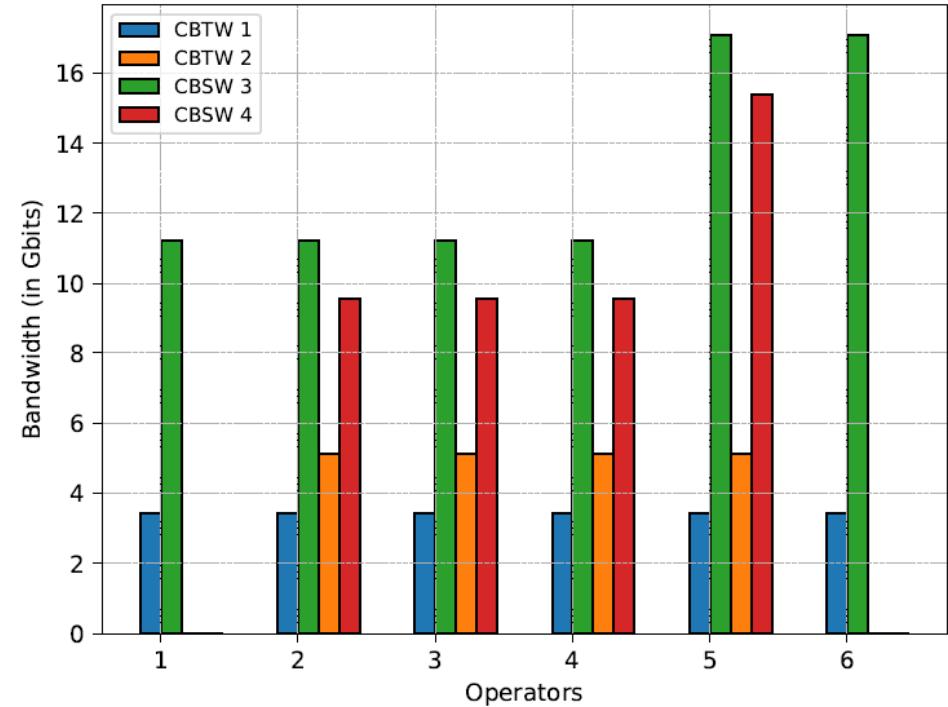
Throughput depends on window semantics

- Load generator is a bottleneck
- Higher parallelization degree and overlap increases bandwidth

Latency

- Independent of count-base vs time-based

Measurement	CBTW	CBSW	TBTW	TBSW
Average Latency	1.76 μ s	1.86 μ s	1.75 μ s	1.87 μ s
Minimum Latency	1.72 μ s	1.8 μ s	1.72 μ s	1.83 μ s
Maximum Latency	1.8 μ s	1.92 μ s	1.8 μ s	1.93 μ s



Stream	$\Sigma = (n, \delta)$	N
CBTW1	$\Sigma = (n = 100, \delta = 100)$	$N = 6$
CBTW2	$\Sigma = (n = 10, \delta = 10)$	$N = 4$
CBSW1	$\Sigma = (n = 5, \delta = 1)$	$N = 6$
CBSW2	$\Sigma = (n = 3, \delta = 1)$	$N = 4$

Some findings

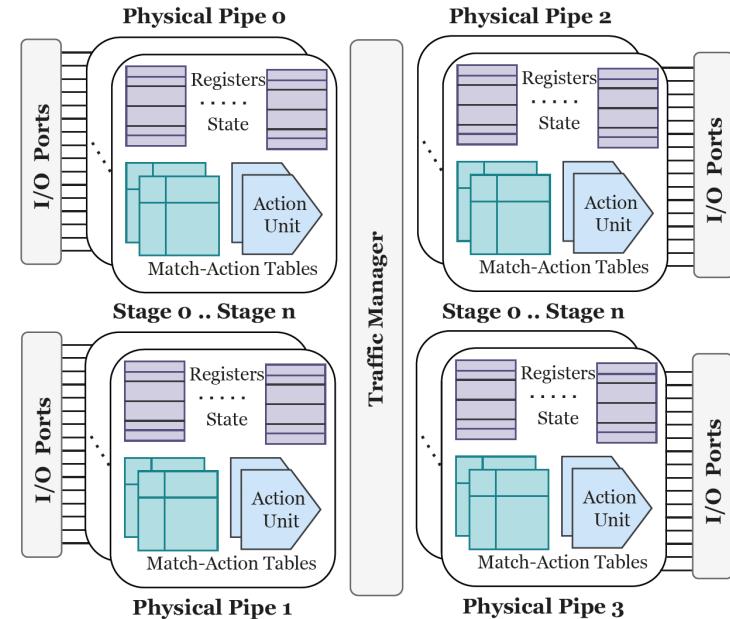
Resource Usage

Resource Usage Determines Scalability

Comprises

- Stages, Tables, and Register Arrays
- Tofino1 has max 12 stages

Resource	CBTW	CBSW	TBTW	TBSW
Stages	6	6	7	8
Match Tables	12	12	18	20
Registers Arrays	3	2	6	7

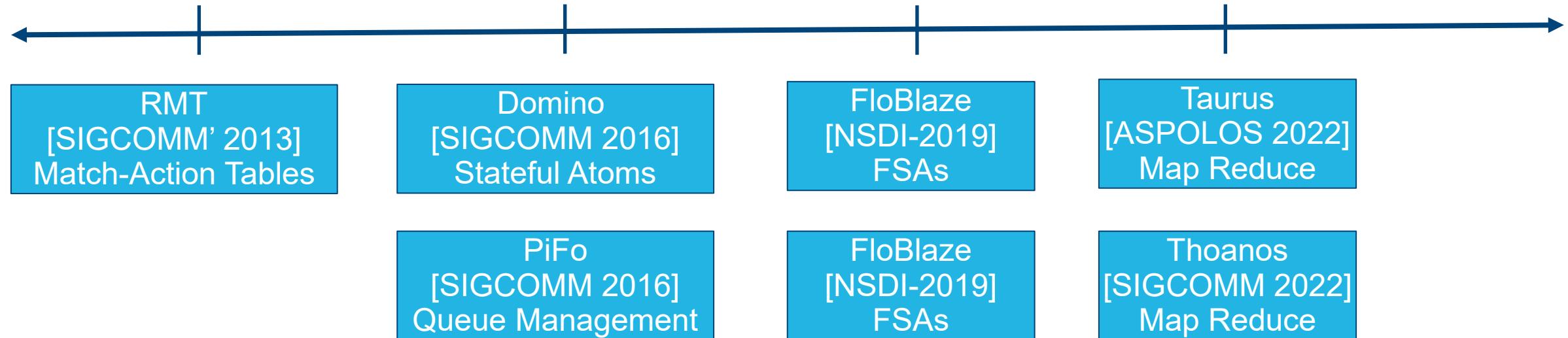


- Could deploy with line rate-performance
 - Count-Based windows : 457k operators, 286k concurrent streams
 - Time-based windows: 362k operators and up to 65k streams

Interesting Approaches in INP for Data Driven Applications

Networking Community is working on many abstractions for Stateful INP

Challenge: understand practicality and applicability in Middleware services



Adapted from Vishal Shrivastav presentation at SIGCOMM

But also very interesting work in distributed computing!

- E.g. “ P4xos: Consensus as a Network Service”, IEEE/ACM Transactions on Networking, 2020.

Everything on Performance?

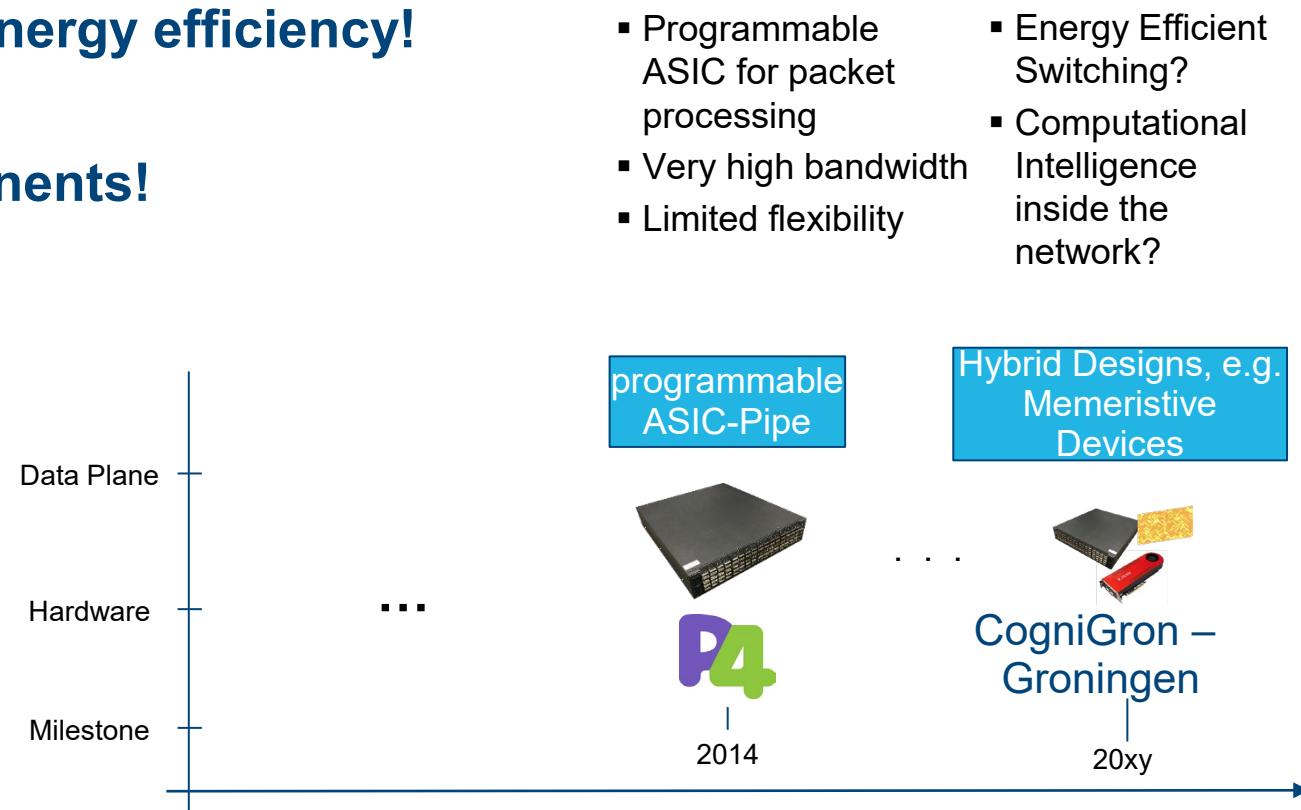
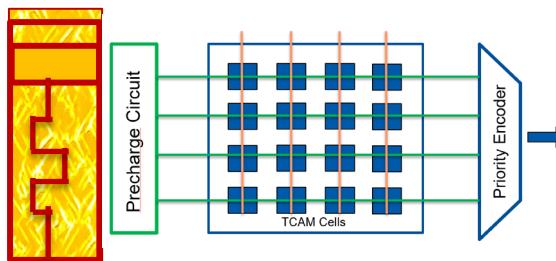
Not really!

Data movements are the cause for high energy efficiency!

Moving to sustainable computing components!

Recent example

- **TCAmM^{CogniGron}: Energy Efficient Memristor-Based TCAM for Match-Action Processing**



ASIC = „fixed silicon chip for special purpose, e.g. packet switching“

Conclusion

Distributed Real-Time Analytics is a fundamental and challenging paradigm in the Internet of Things

Accelerators based on In-Network Computing

- Reduce performance bottlenecks
- Utilize the Distributed Infrastructure more efficient

Distributed systems mechanisms

- Flexible usage of heterogeneous resources
- No single mechanism fits them all

Future Research:

- Better understanding of
Distributed Computing + In-Network computing
- Energy-efficiency of In-Network Computing

Questions



Ralf Kundel and Fridolin Siegmund and Rhaban Hark and Amr Rizk and Boris Koldehofe. Network Testing Utilizing Programmable Networking Hardware. IEEE Communications Magazine, 7 pages, IEEE 2022.

Bowmik, Tariq, Koldehofe, Kohler, Dürr, Rothermel. High Performance Publish/Subscribe Middleware in Software-defined Networks. IEEE Transactions on Networking (ToN), 2016.

Bochra Boughzala, Christoph Gärtner, and Boris Koldehofe. Window-based Parallel Operator Execution with In-Network Computing. Proceedings of the 16th ACM International Conference on Distributed and Event-based Systems (DEBS '22), pp. 91–96, ACM press.

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