STRETCH: Scalable and Elastic Deterministic Streaming Analysis with Virtual Shared-Nothing Parallelism

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Distributed Computing and Systems Chalmers university of technology



Improve performance by:

- Scale Down the amount of data (computing resources)
- Scale Up the computing resources on a node via parallel processing
- Scale Out the computing to distributed nodes

Phillip B. Gibbons, Keynote Talk IPDPS'15

Big Data: Scale Down, Scale Up, Scale Out

Scale Up before Scale Out

Scale Up

Scale Out

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Big Data: Scale Down, Scale Up, Scale Out

Scale Up before Scale Out

- Often order of magnitude better performance if data fits in memory of multicore
- Multicores have 1-12 TB memory
- Even when data doesn't fit, will still want to take advantage of Scale Up whenever you can

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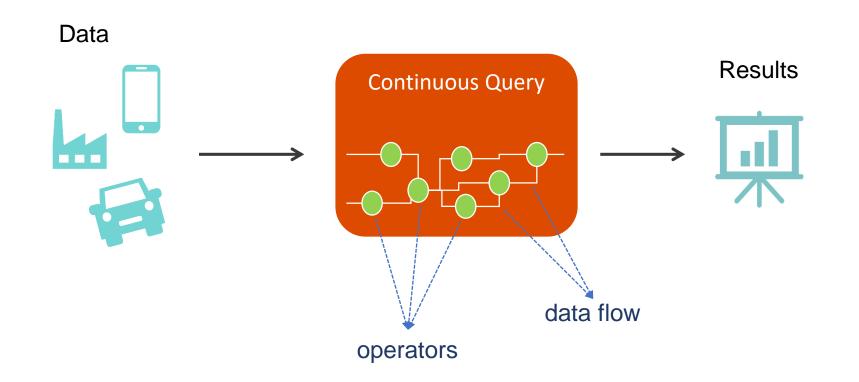
Big Data: Scale Down, Scale Up, Scale Out

Adjusting resources on node level for stateful streaming analysis

What is stream processing?

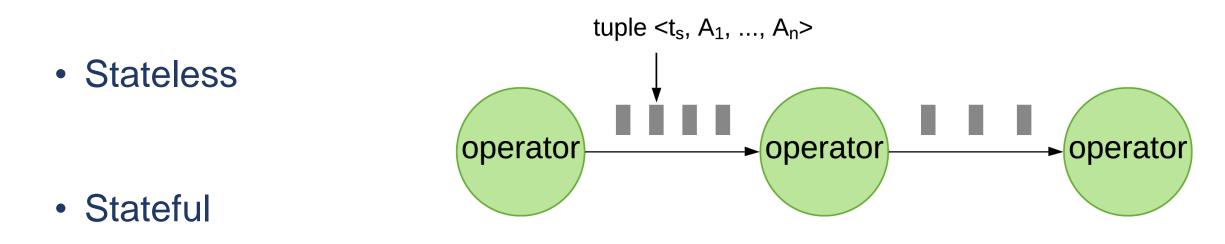


Data stream processing



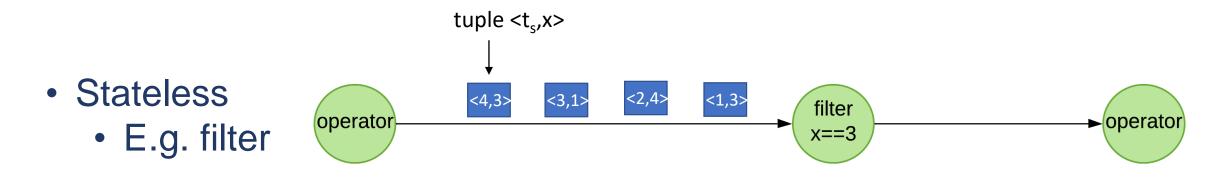
Stream Processing Operators

State is the memory of the operator



Stream Processing Operators

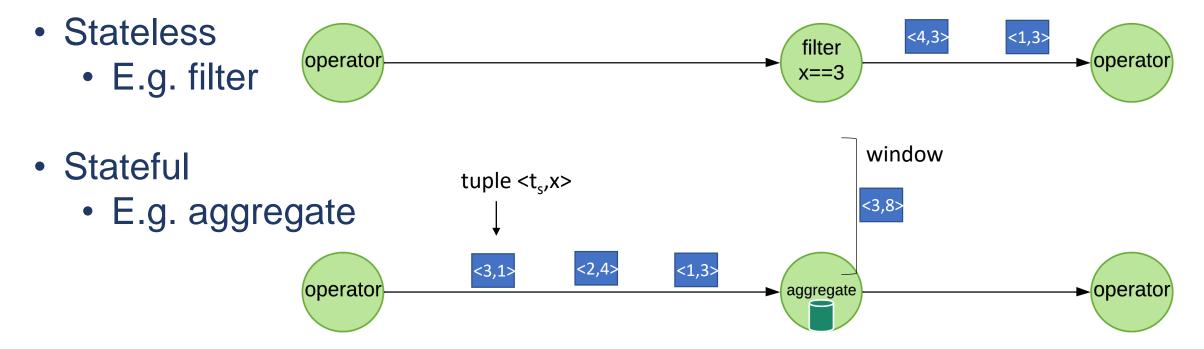
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• Stateful

Stream Processing Operators

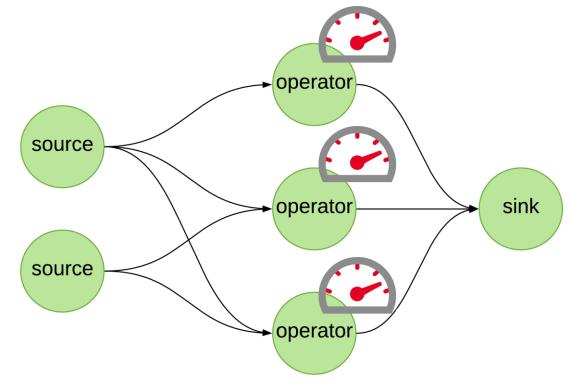
State is the memory of the operator



Stream Processing Performance

• Throughput

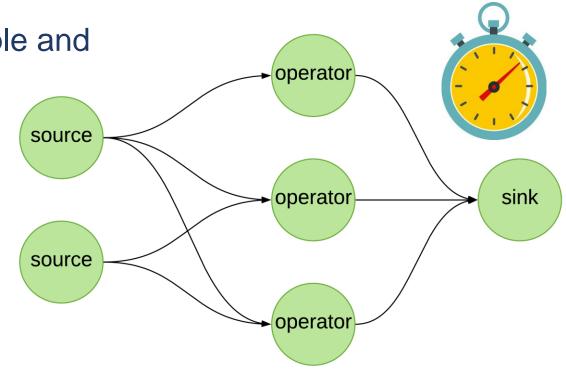
Number of tuples processed per time unit



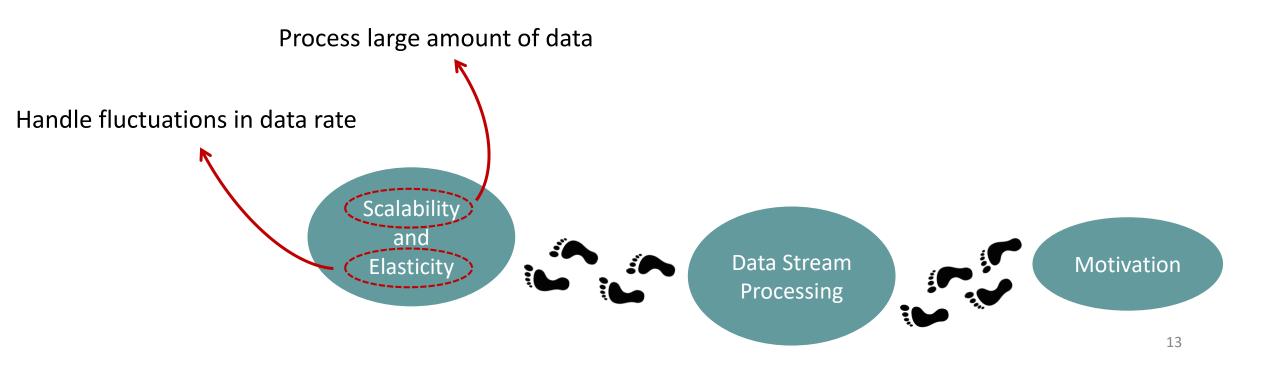
Stream Processing Performance

- Throughput
- Latency

Time difference between receiving a tuple and producing the corresponding results

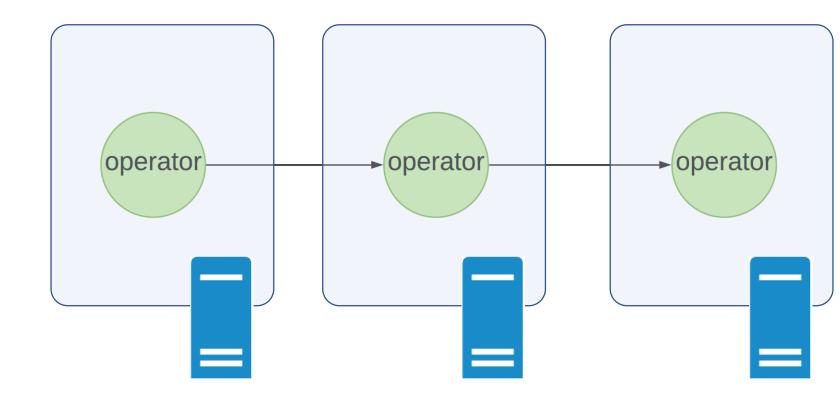


Challenges



Stream Processing Scalability

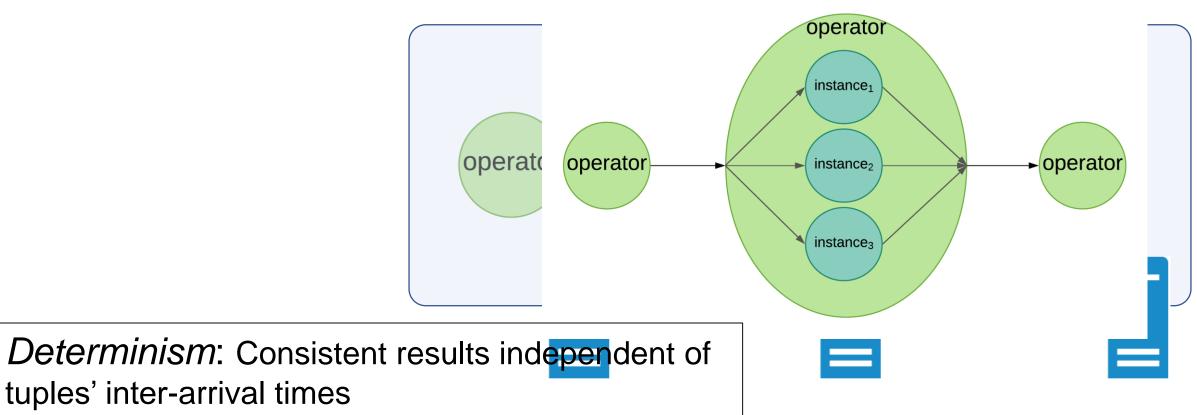
• Pipeline parallelism



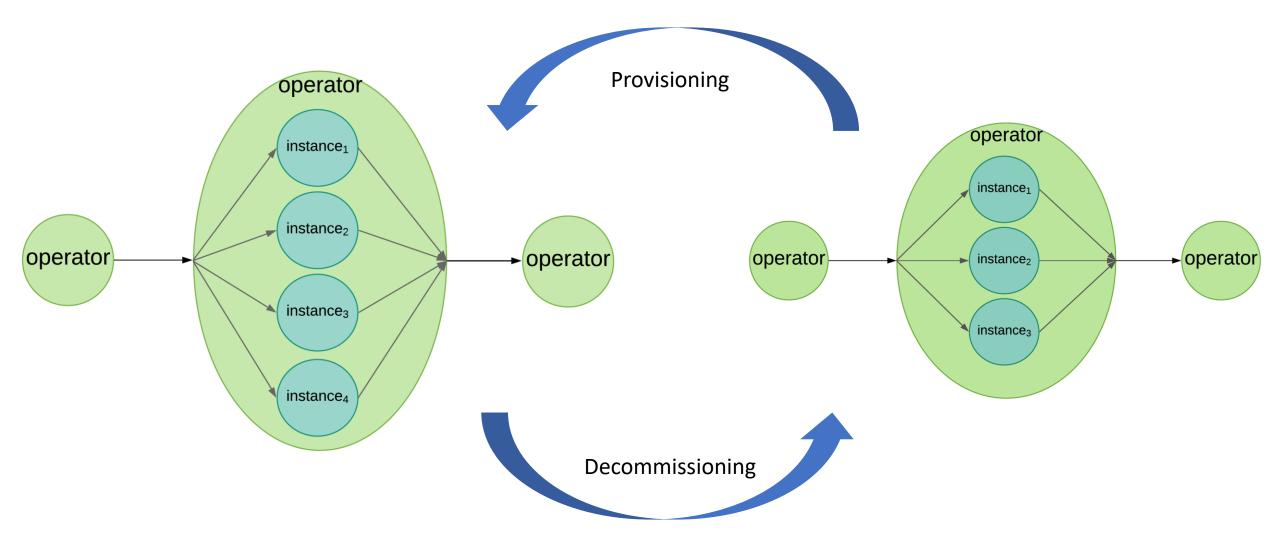
Stream Processing Scalability

• Pipeline parallelism

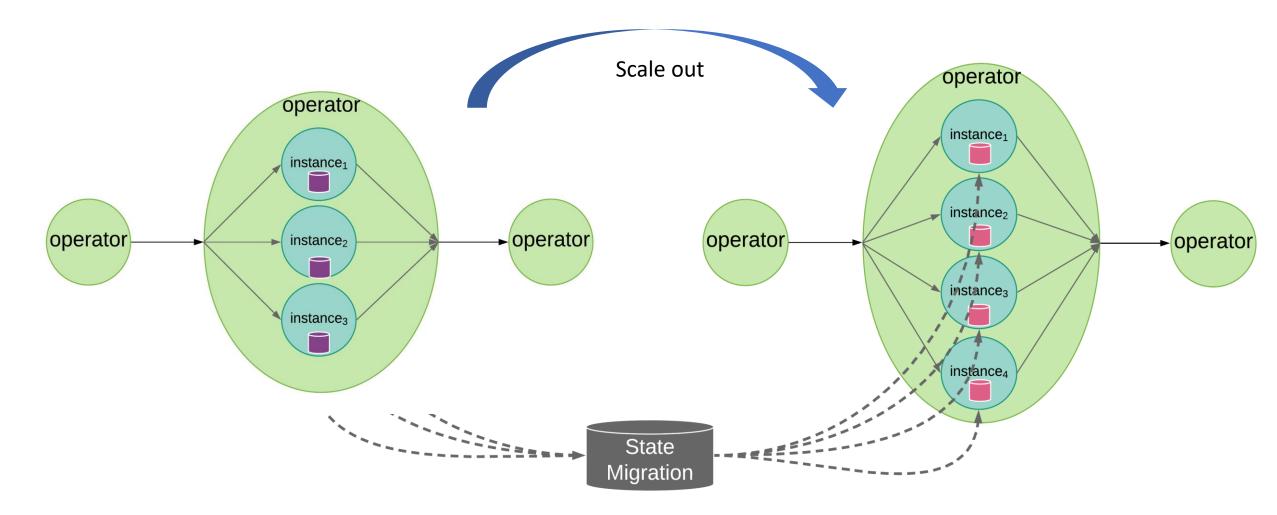
• Data parallelism



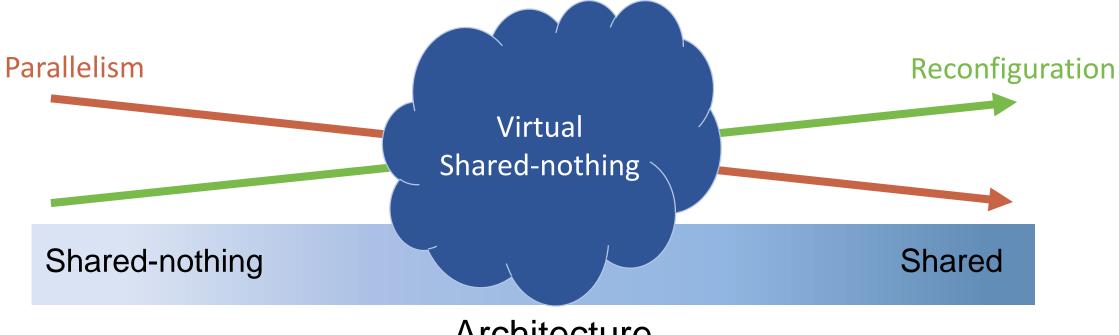
Stream Processing Elasticity



Stream Processing Elasticity

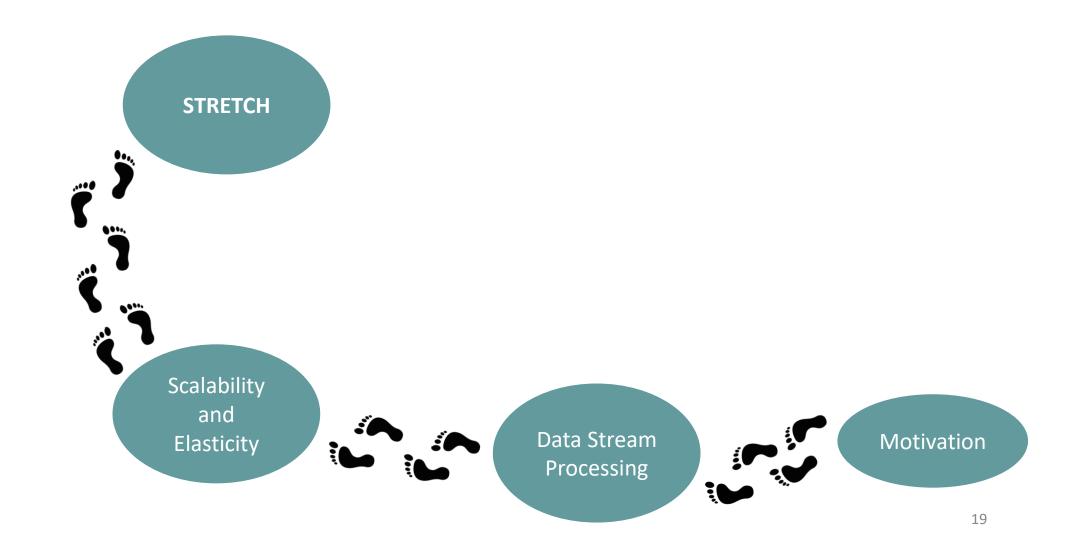


Stream Processing Efficiency



Architecture

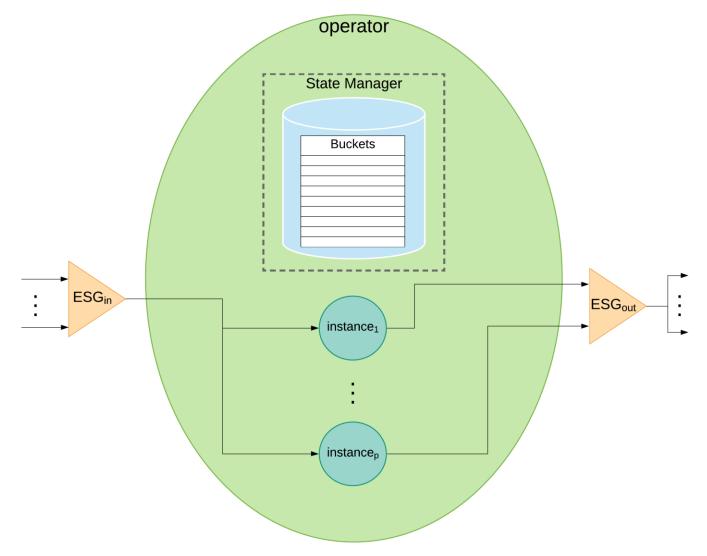
Proposed Framework



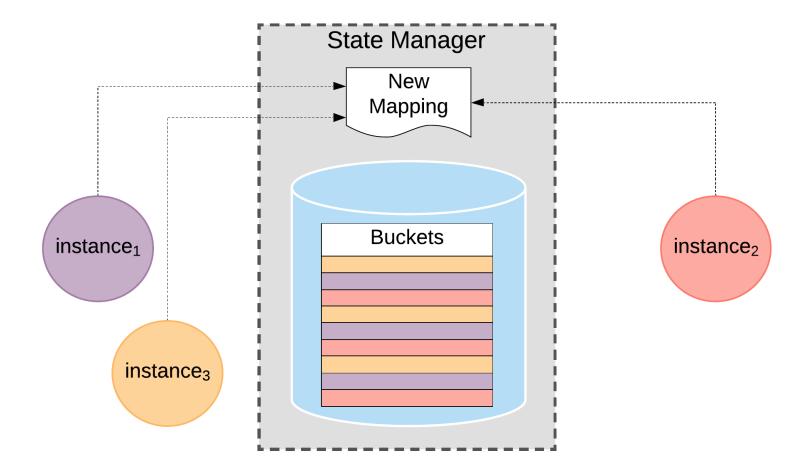
STRETCH Framework

Components:

- State manager
 - Virtual shared-nothing parallelism



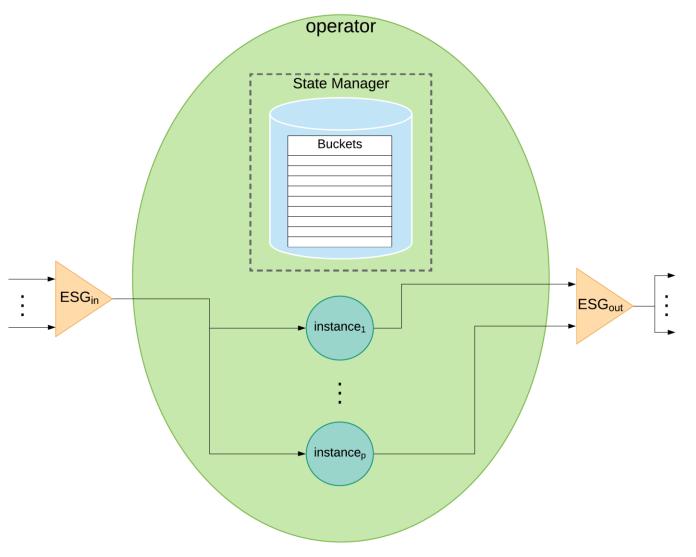
Virtual Shared-nothing Parallelism



STRETCH Framework

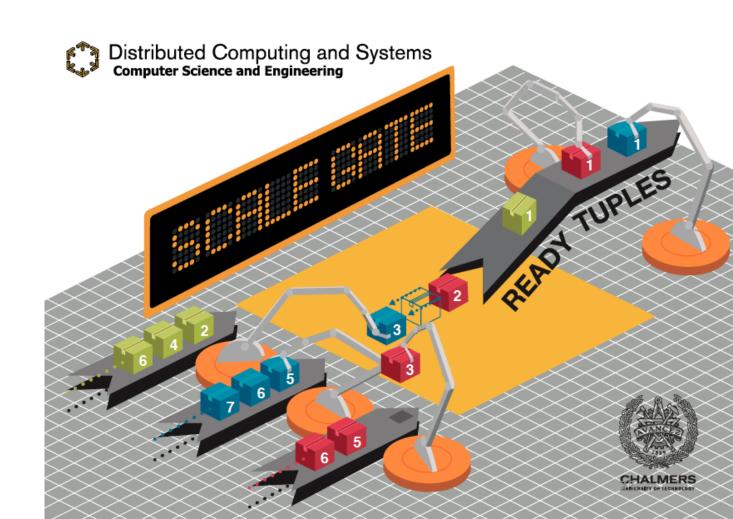
Components:

- State manager
 - Virtual shared-nothing parallelism
- Elastic ScaleGate (ESG)



ScaleGate

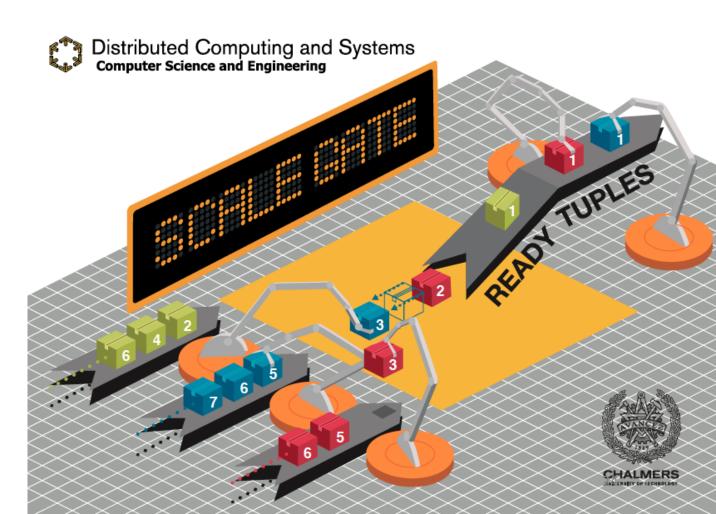
- Sort concurrent arriving tuples based on timestamp
- Lock-free data structure

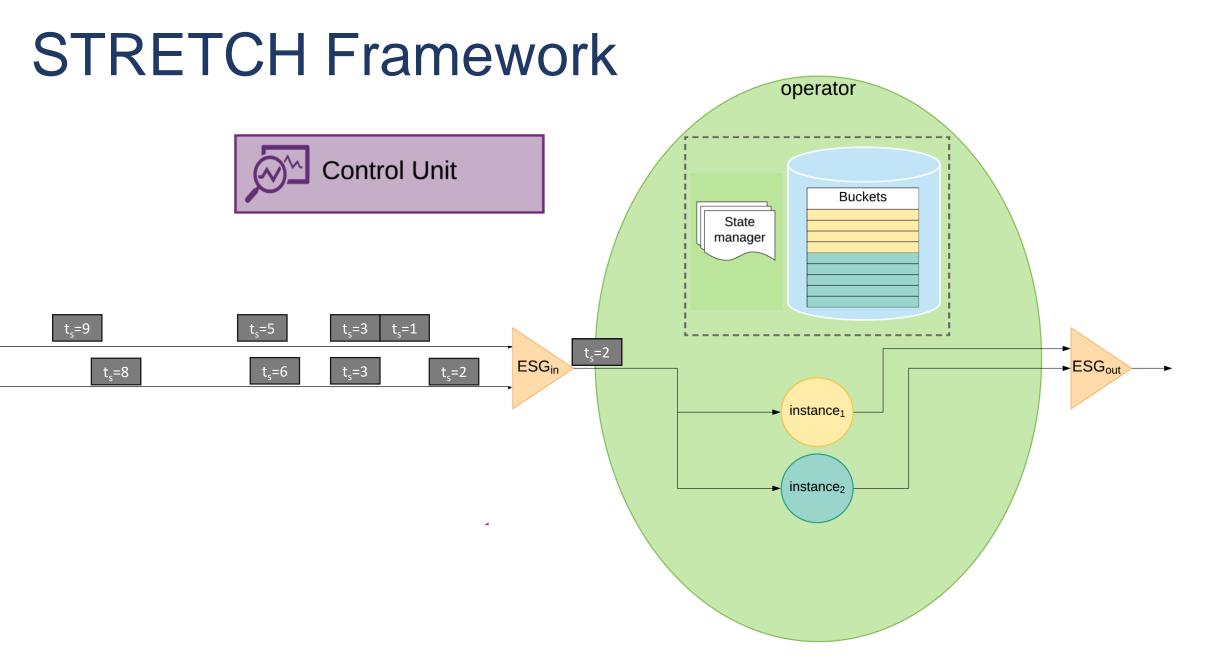


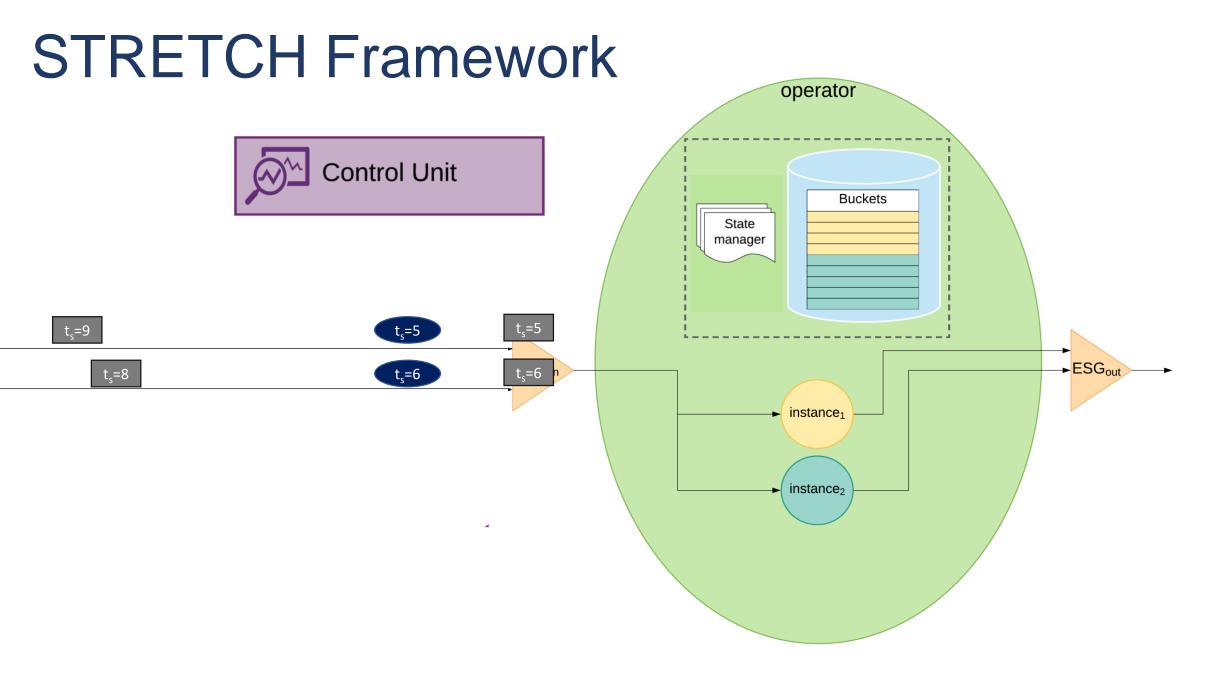
Stastic Steale Gate

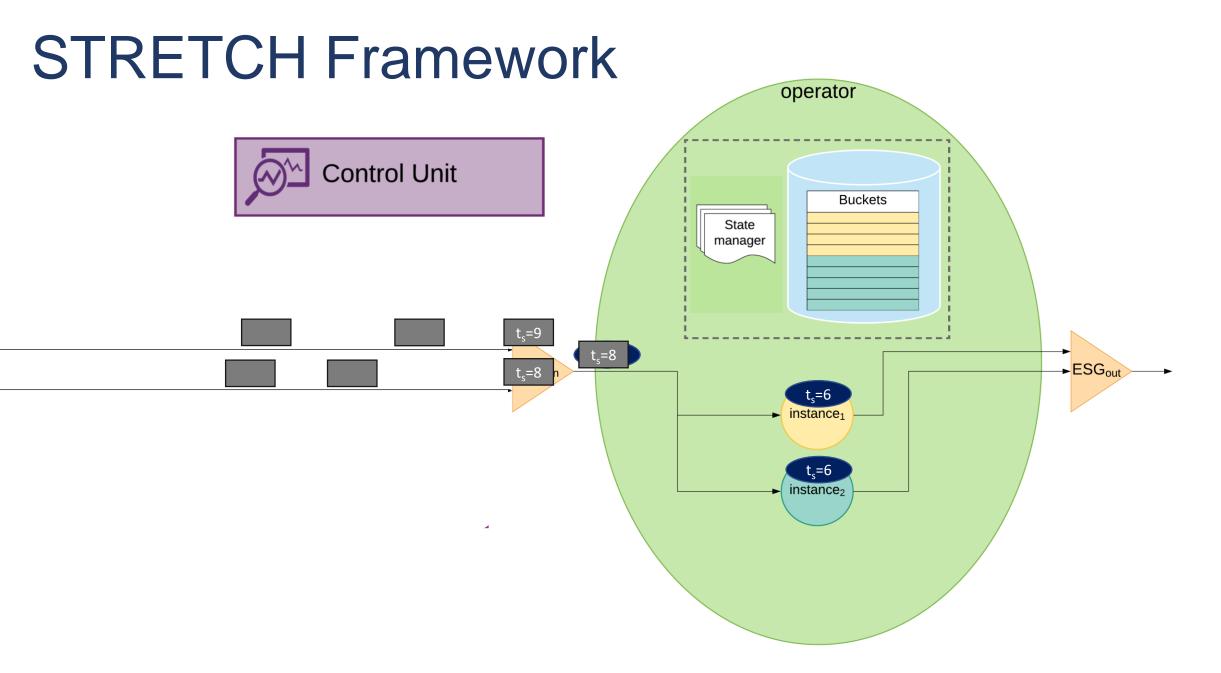
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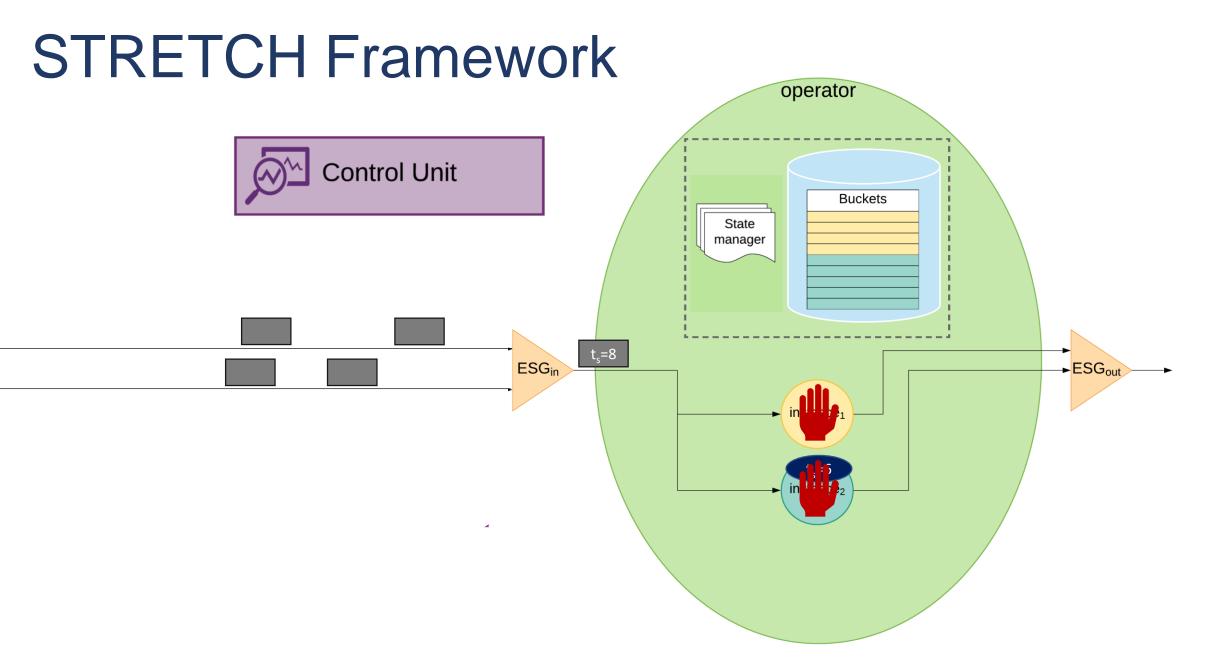
 Changing number of readers/sources at runtime

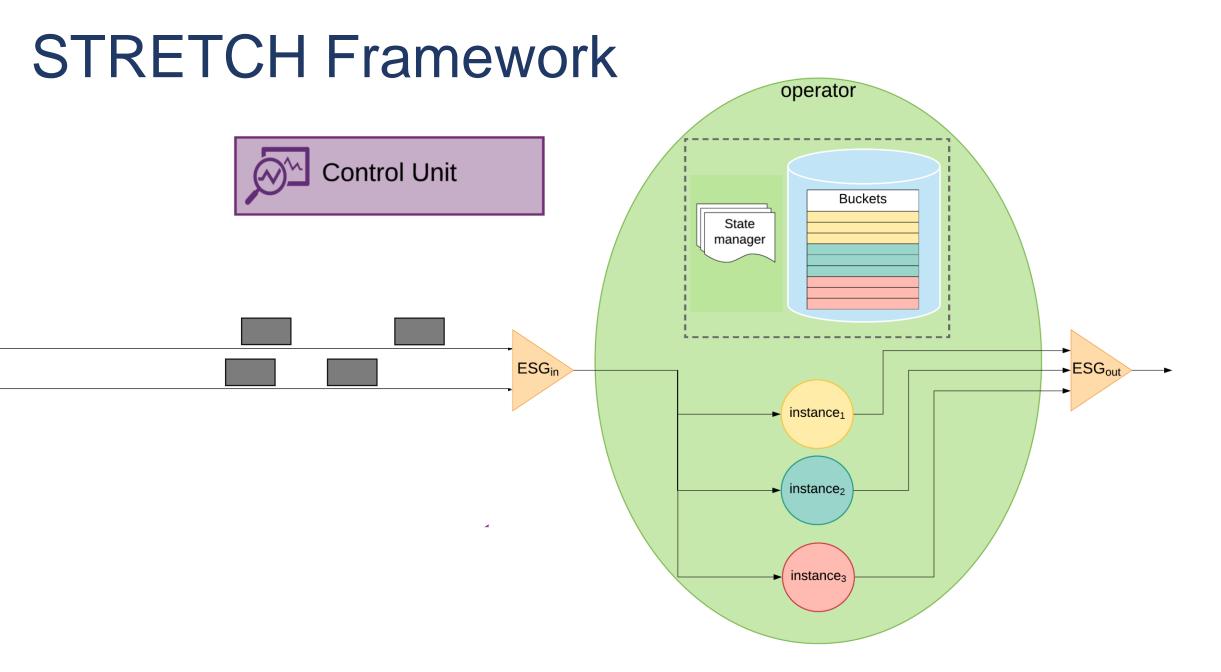












Performance Evaluation

- Setup: Intel Xeon E5-2695
- Use case: ScaleJoin

t₅

t₄

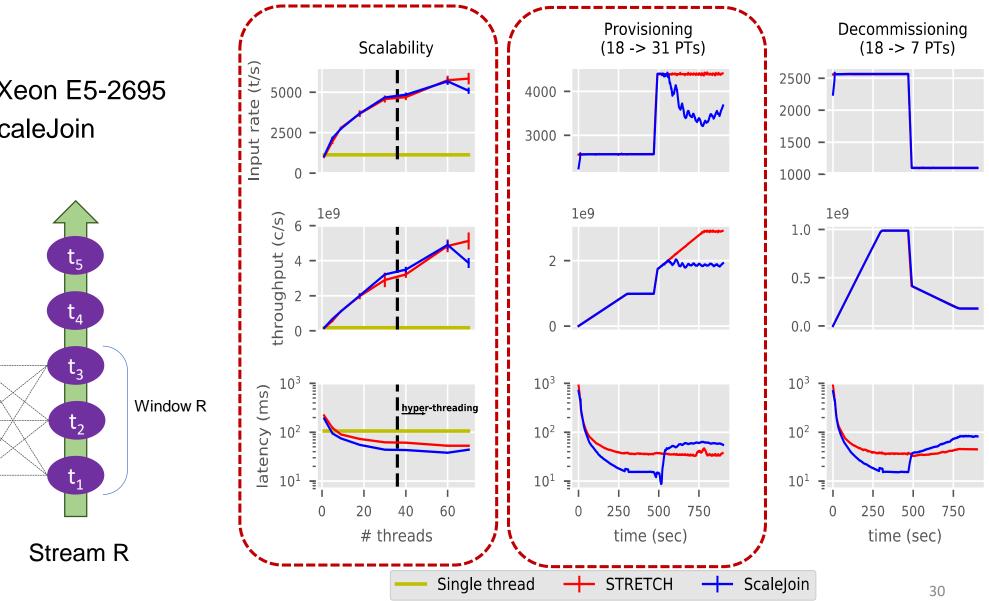
 t_3

 t_2

t₁

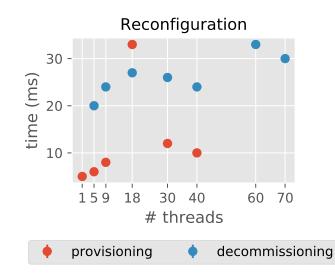
Stream S

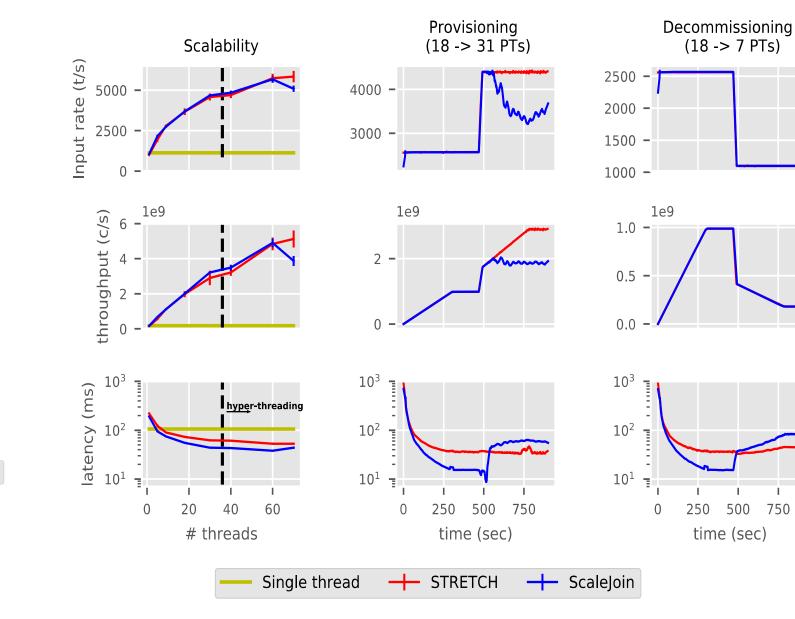
Window S

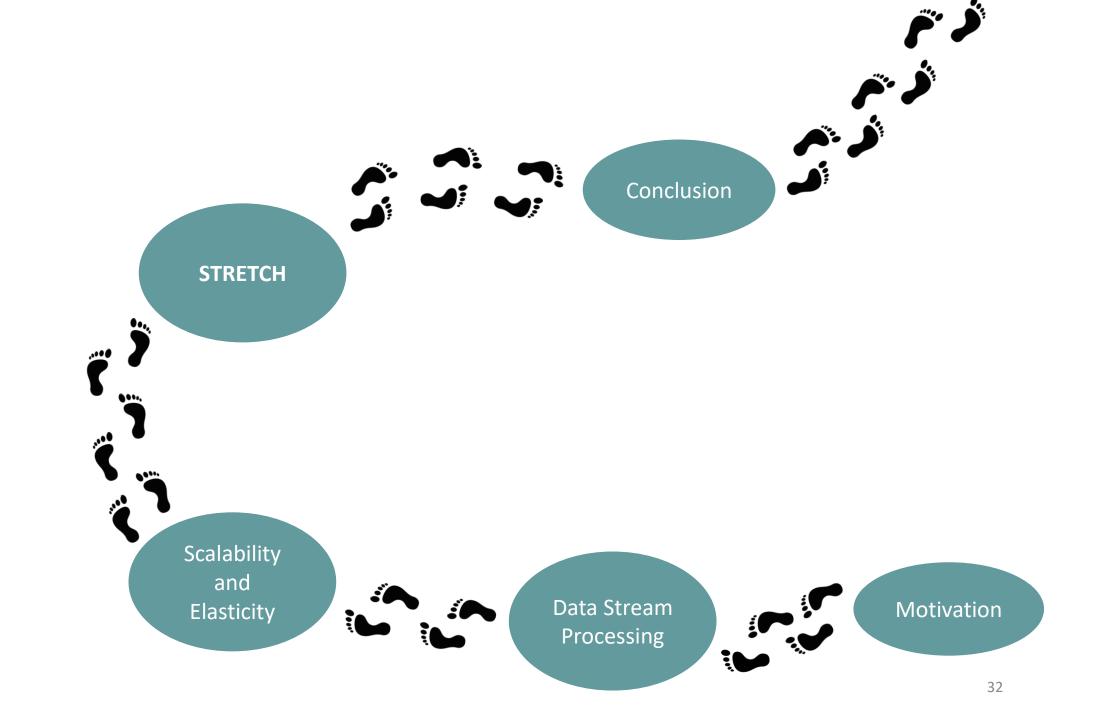


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Conclusion

- Virtual shared-nothing parallelism
- Adaptive reconfiguration of processing units
- Intra-node resource utilization
- Deterministic execution

- Scale up/scale out
- Automatic control unit





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