

# SnailTrail

# Generalizing Critical Paths for Online Analysis of Distributed Dataflows

**Moritz Hoffmann**, Andrea Lattuada, John Liagouris, Vasiliki Kalavri, Desislava Dimitrova, Sebastian Wicki, Zaheer Chothia, and Timothy Roscoe

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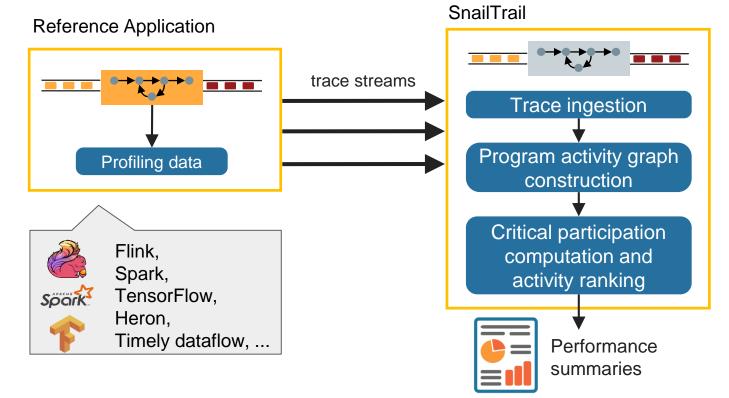




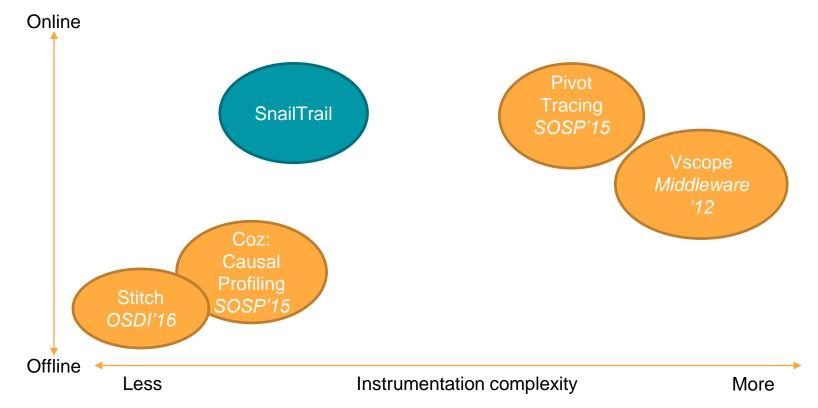
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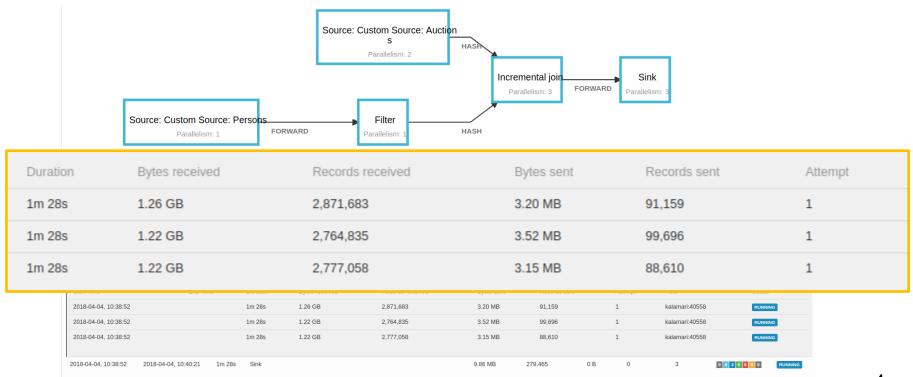
# SnailTrail: Diagnosing latency issues in dataflows *"Where is the latency bottleneck in my computation?"*



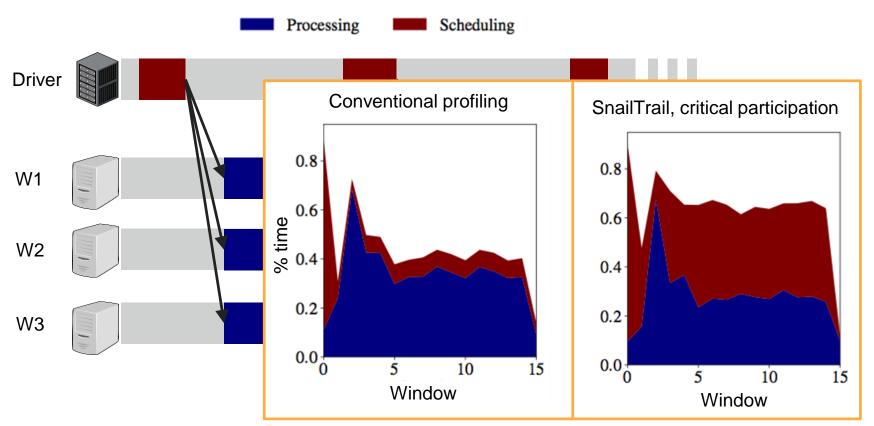
## SnailTrail works online with minimal instrumentation



#### Example 1: Metrics in Flink's dashboard



## Example 2: Task Scheduling in Spark



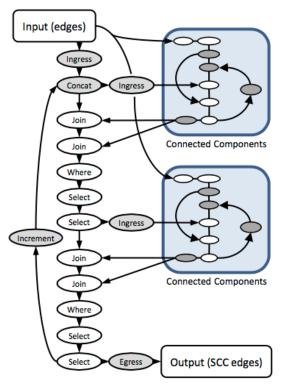
Spark

#### The real-world is more complex

Many tasks, activities, operators, dependencies

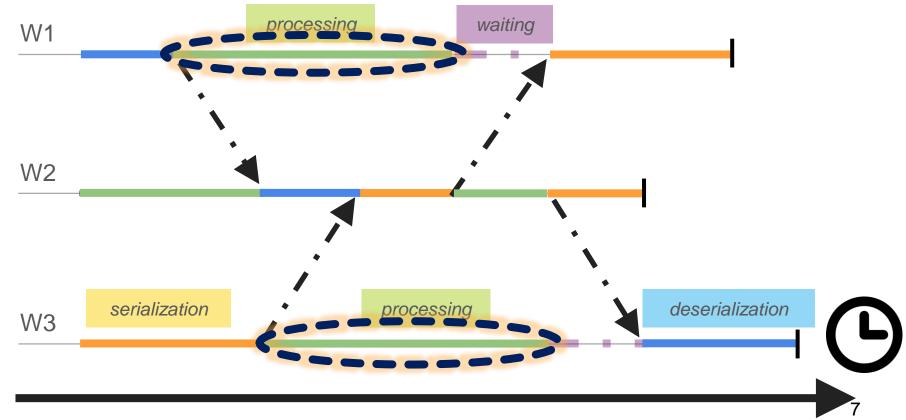
Long-running, dynamic workloads

Bottlenecks not isolated

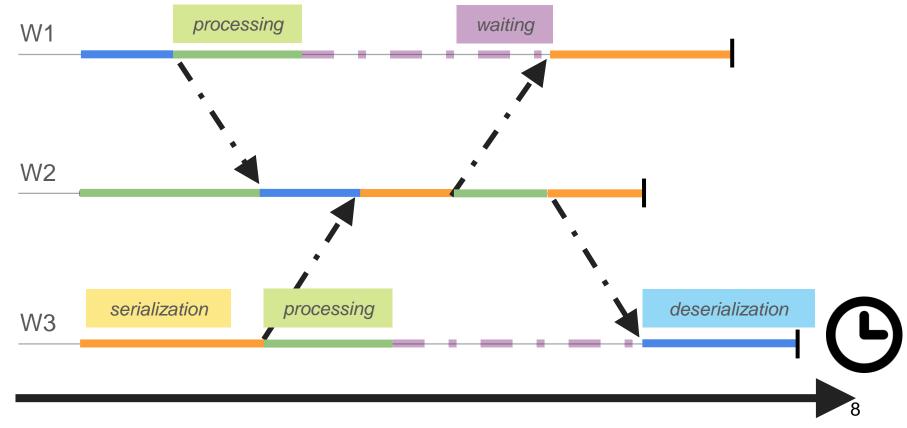


Credits: Frank McSherry, "Tracking progress in timely dataflow"

#### Conventional profiling can indicate wrong bottleneck

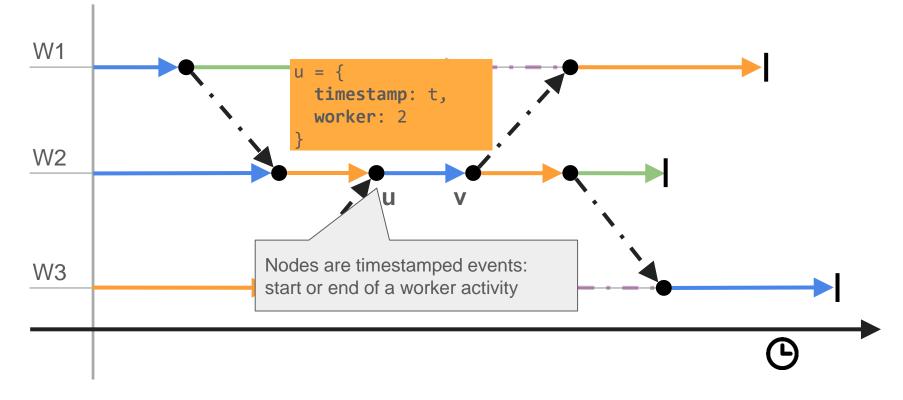


#### Conventional profiling can indicate wrong bottleneck

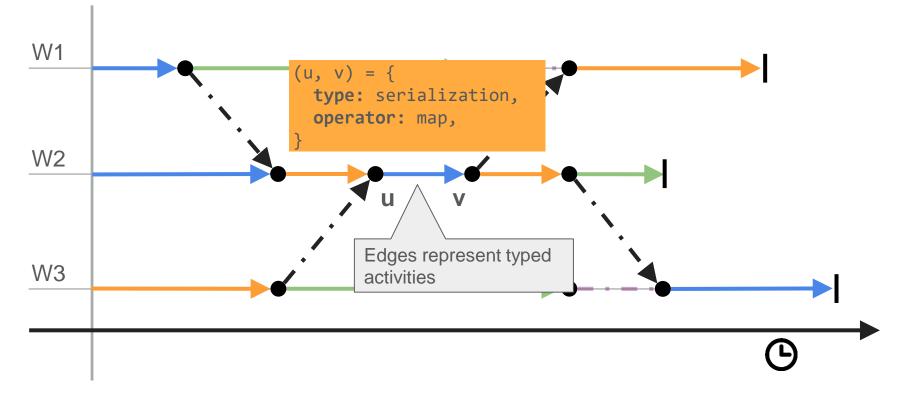


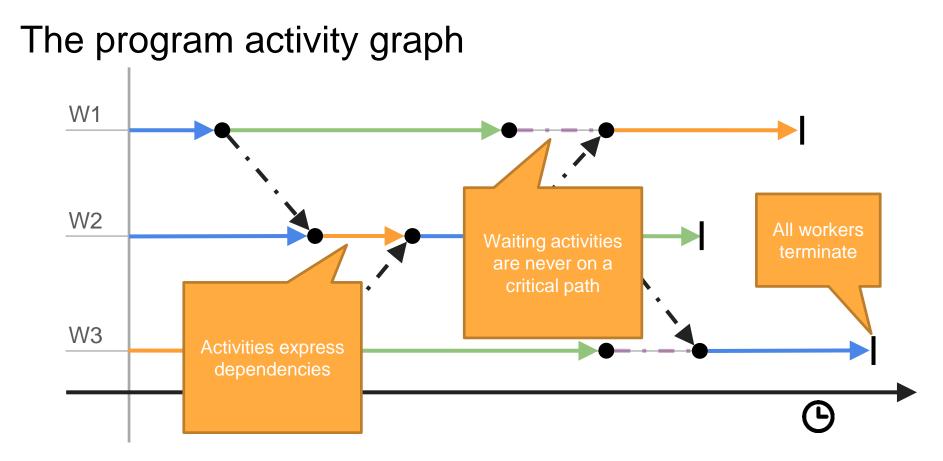
A quick review of critical path analysis

## The program activity graph



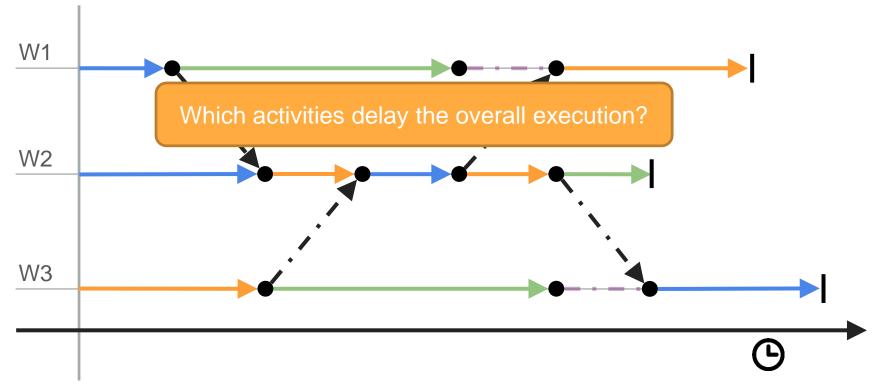
## The program activity graph



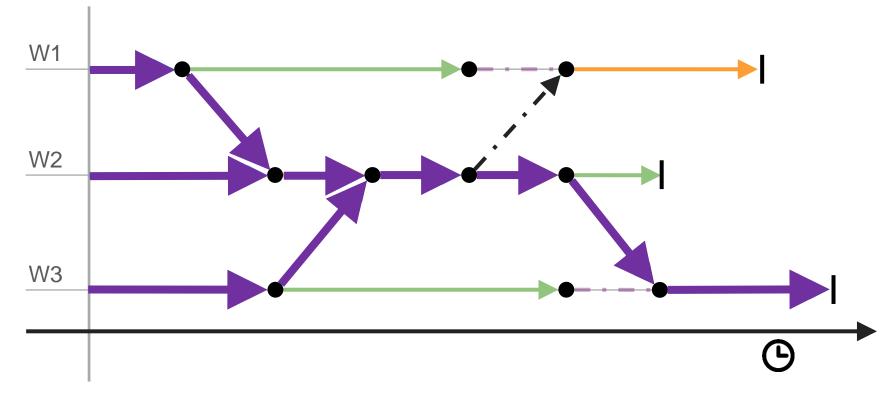


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### The program activity graph



#### Classical critical path analysis



What is the equivalent of a critical path for continuously running, distributed streaming applications, with potentially **unbounded** input?

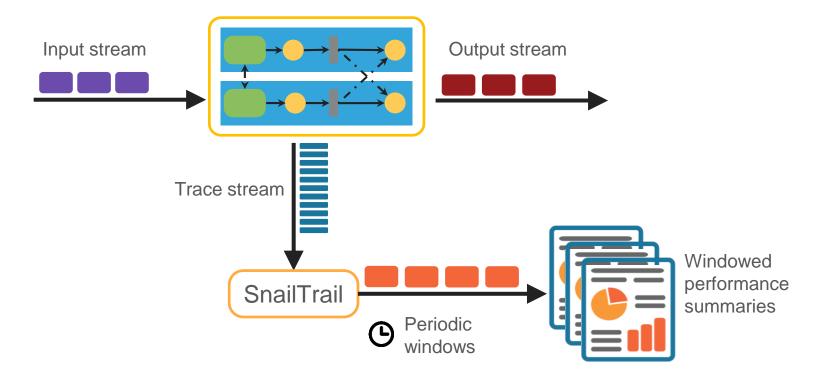
There might be no "job end"

The program activity graph and critical paths change continuously

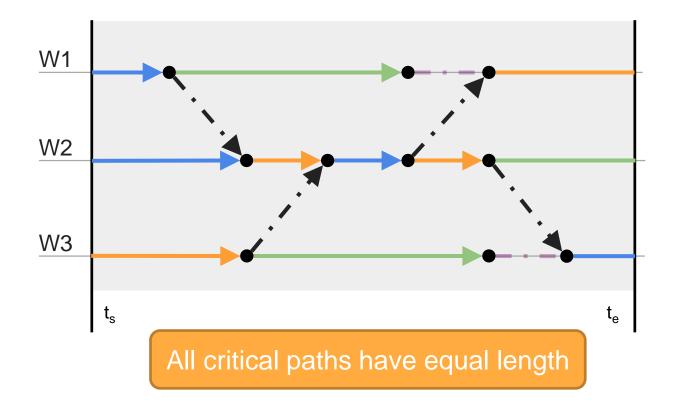
Profiling information can quickly become stale

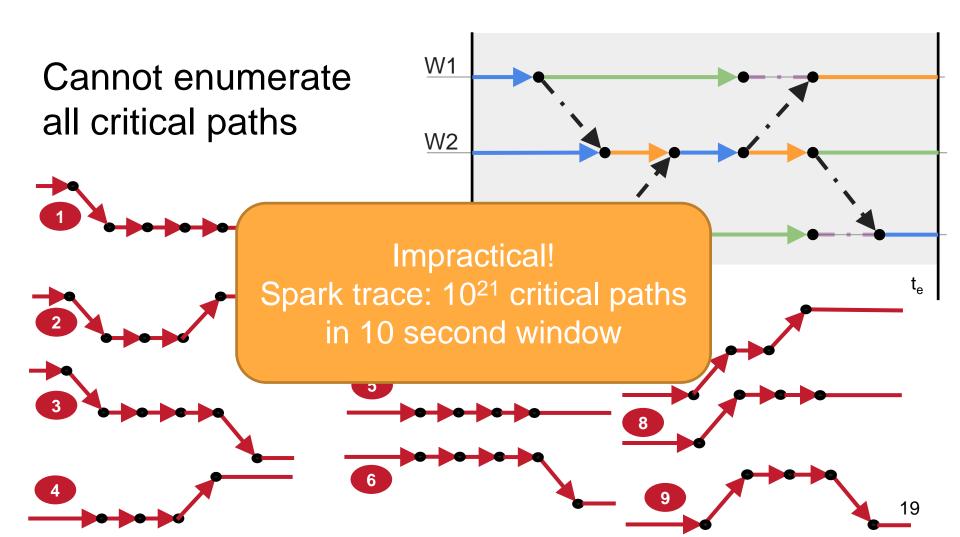
# Online critical path analysis

## SnailTrail: Online analysis of trace windows

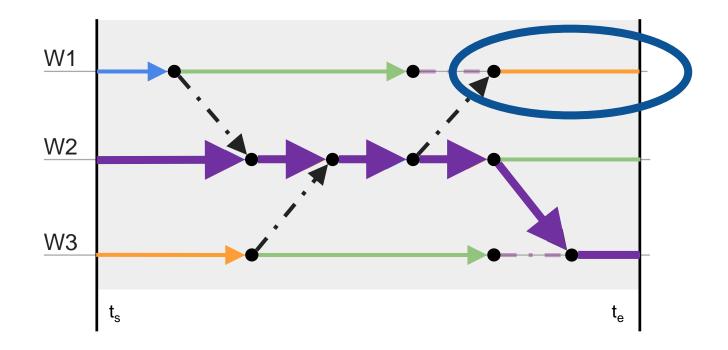


#### Program activity graph window





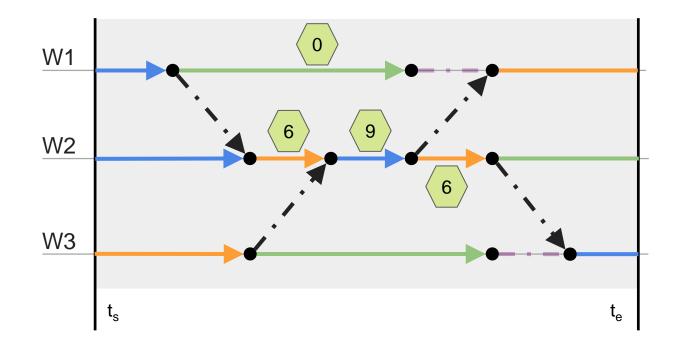
#### Sampling critical paths misses critical activities



We **rank activities** across all critical paths to capture their relative importance.

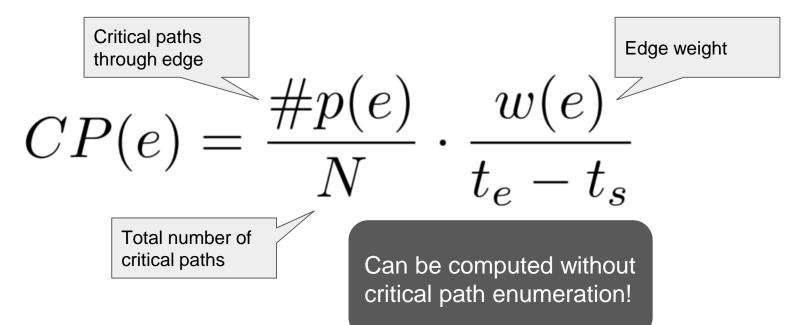
Intuition: The more critical paths go through an activity, the more critical it might be

## Counting over enumerating

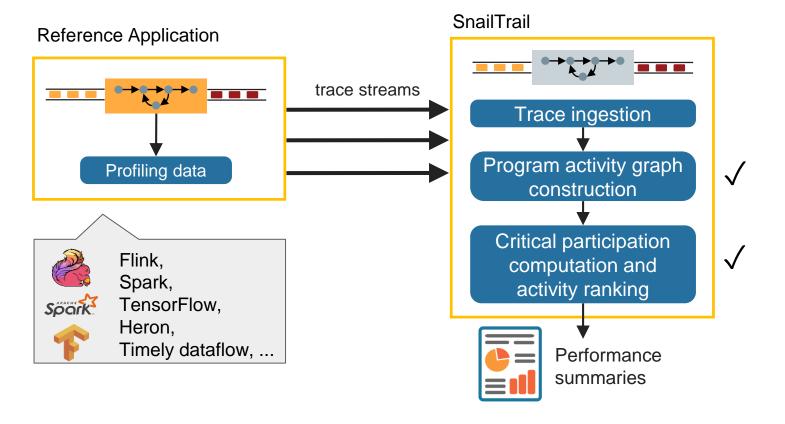


### The Critical Participation metric

Fraction of an edge's time contribution across all critical paths

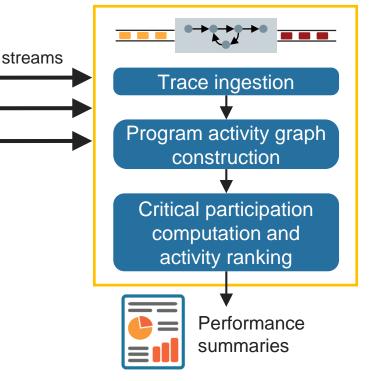


# SnailTrail in action



## Interpreting critical participation-based summaries

SnailTrail



Stream of tuples:

(Activity type, Operator, Worker, ..., Critical participation)

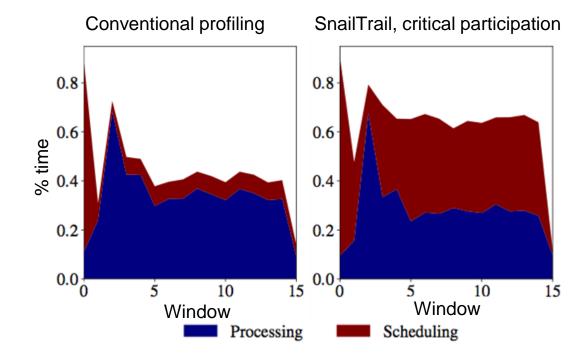
Examples:

Activity type bottleneck analysis Operator bottleneck analysis

(More in the paper!)

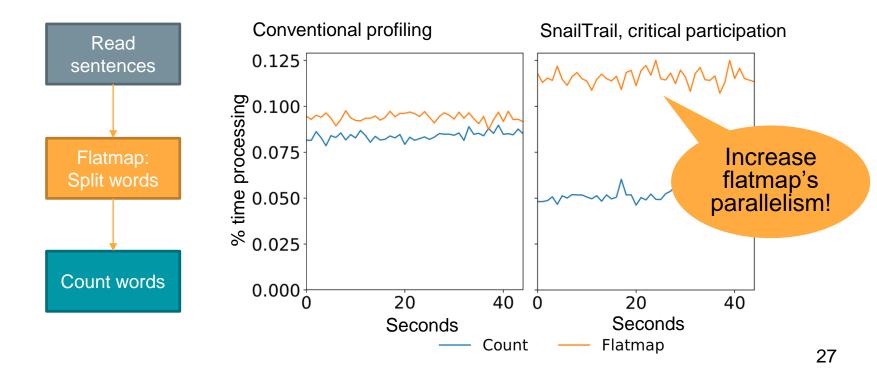
#### Activity type bottleneck analysis (Spark)

Apache Spark: Yahoo! Streaming Benchmark, 16 workers, 8s windows



### Operator bottleneck analysis (Flink)

Apache Flink: Dhalion WordCount Benchmark, 10 workers, 1s windows



# SnailTrail performance

Low instrumentation overhead

Spark, TensorFlow No observed overhead Flink, Timely ~10% overhead compared to

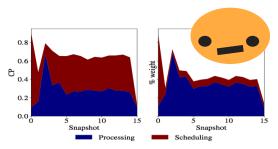
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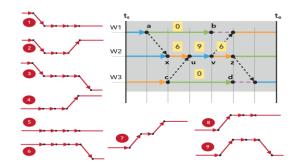
High throughput 1.2 million events/s 8 workers Always online 1s of traces in 6ms (100x) 256s of traces in < 25s (10x)

SnailTrail on Intel Xeon E5-4640, 2.40GHz, 32 cores, 512GiB RAM Trace: Apache Flink Sessionization, 48 workers, 1s-256s windows

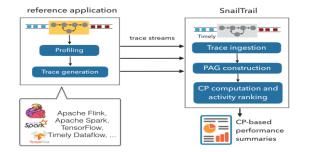
# Summary



#### Conventional profiling is misleading



#### CP-metric: online critical path analysis



SnailTrail: online CP-based summaries

