

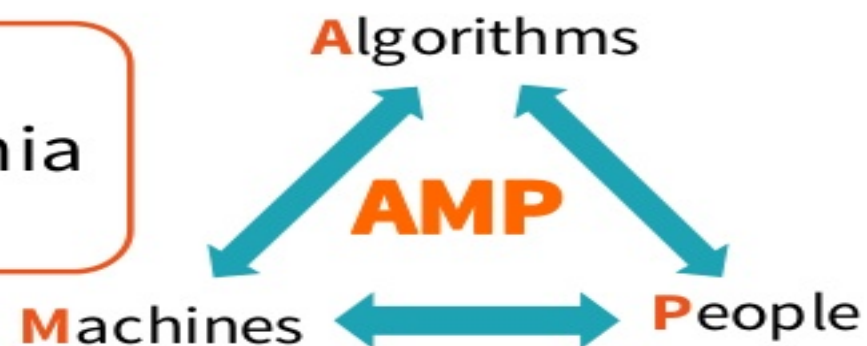
RISELab: Enabling Intelligent Real-time Decisions

Ion Stoica
February 8, 2017



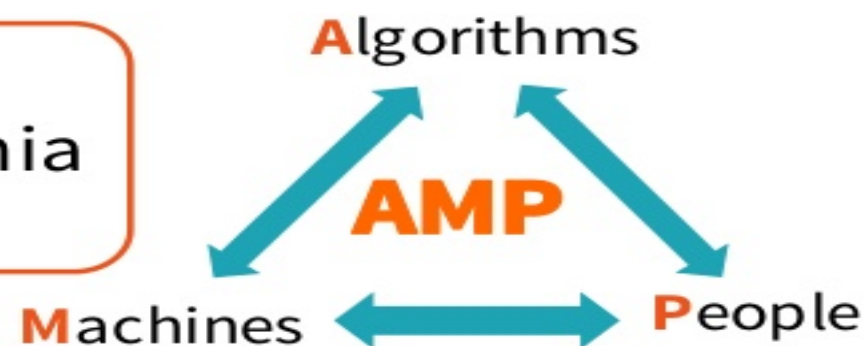
Berkeley's AMPLab (2011-2016)

Goal: Next generation of open source data analytics stack for industry & academia
Berkeley Data Analytics Stack (BDAS)



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Apache
MESOS™



ALLUXIO ...

RISE: Real-time Intelligent Secure Execution



RISELab

From **live data** to **real-time decisions**



AMPLLab

From **batch data** to **advanced analytics**

RISE Lab (2017-2022)



12 faculty across AI, systems, security, and architectures



11 Founding sponsors



Why?

Data only as valuable as the **decisions** it enables



How to Turn Big Data into Insights and Action

by Bernard Marr | May 18, 2016 6:30 am | 1 Comments



Forrester's 2016 Predictions: Turn Data Into Insight And Action

Posted by [Brian Hopkins](#) on November 9, 2015



Why?

Data only as valuable as the **decisions** it enables

What does this mean?

- **Faster** decisions better than slower decisions
- Decisions on **fresh** data better than decisions on stale data
- Decisions on **personalized** data better than on aggregate data

Goal

Real-time decisions

decide in ms

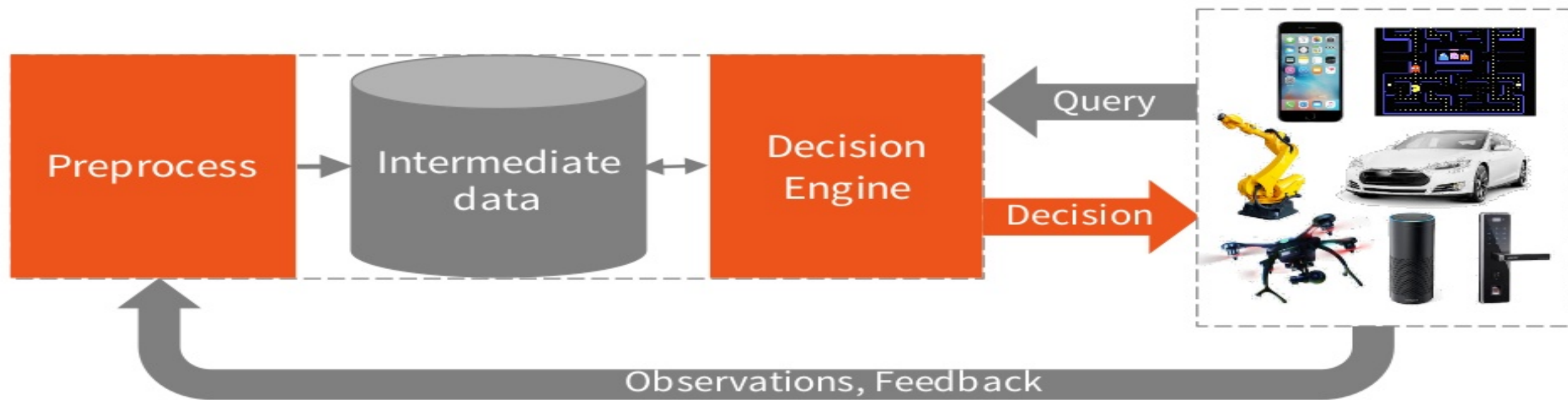
on live data

the current state of the environment

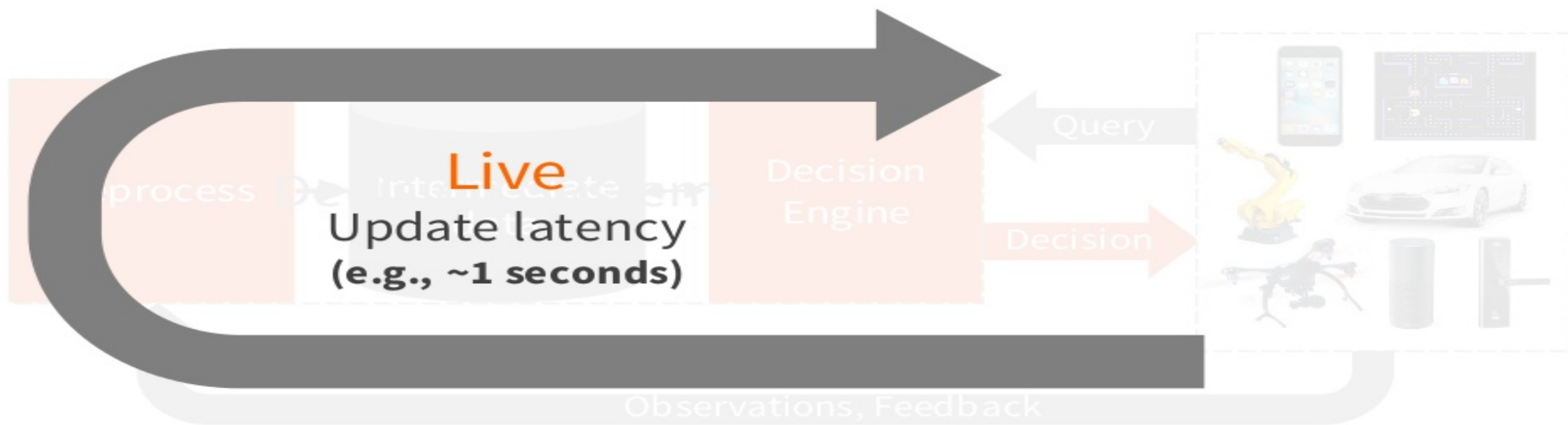
with strong security

privacy, confidentiality, integrity

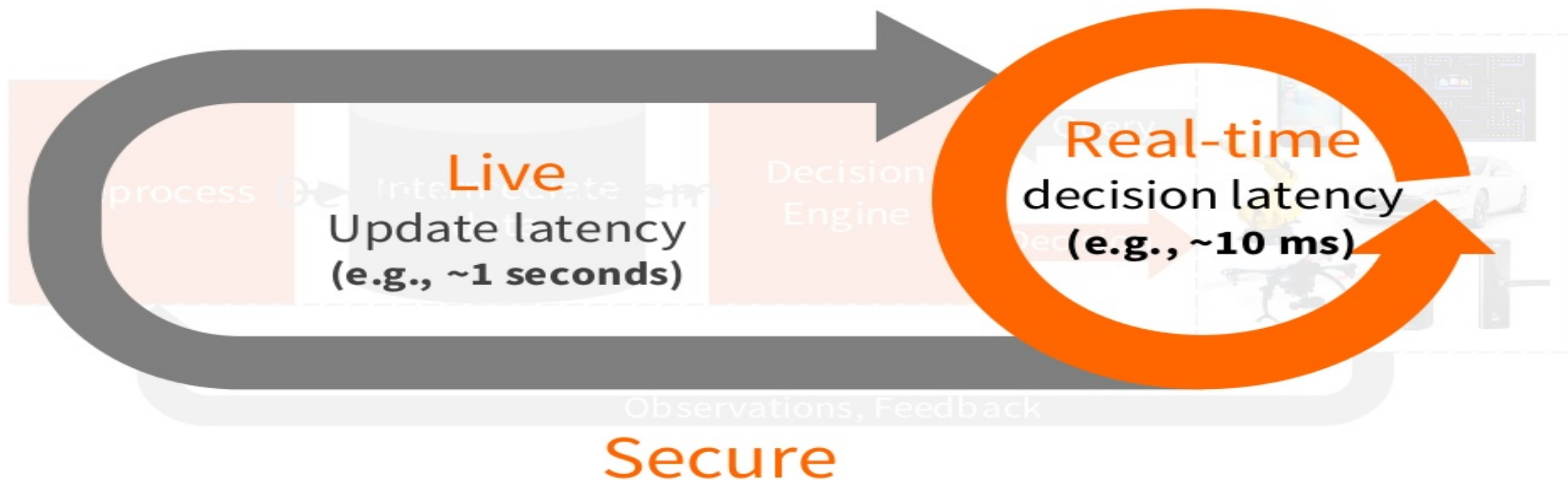
Typical decision system



Typical decision system

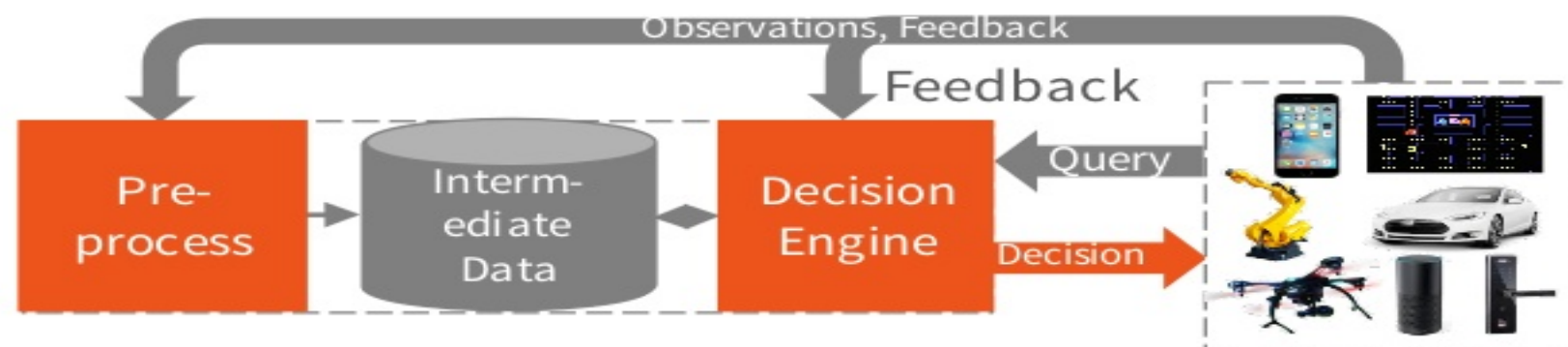


Typical decision system



Example of decision systems

ML Pipeline
(e.g., Clipper +
Spark/Tensorflow)



Reinforcement
Learning Systems
(e.g., Ray)



What else do we want from decisions?

Intelligent: complex decisions in uncertain environments

Robust: handle complex noise, unforeseen inputs, failures

Explainable: ability to explain non-obvious decisions

Goal

Develop **open source**
platforms, tools, and algorithms for
intelligent real-time decisions on live-data

Some Proposed Research

Secure Real-time Decisions Stack (SRDS)

- Open source platform to develop of RISE apps
- Secure from ground up
- Reinforcement Learning (RL) as one of key app patterns

Learning control hierarchies: speedup learning, training

Shared learning: learn over confidential data

Some Proposed Research

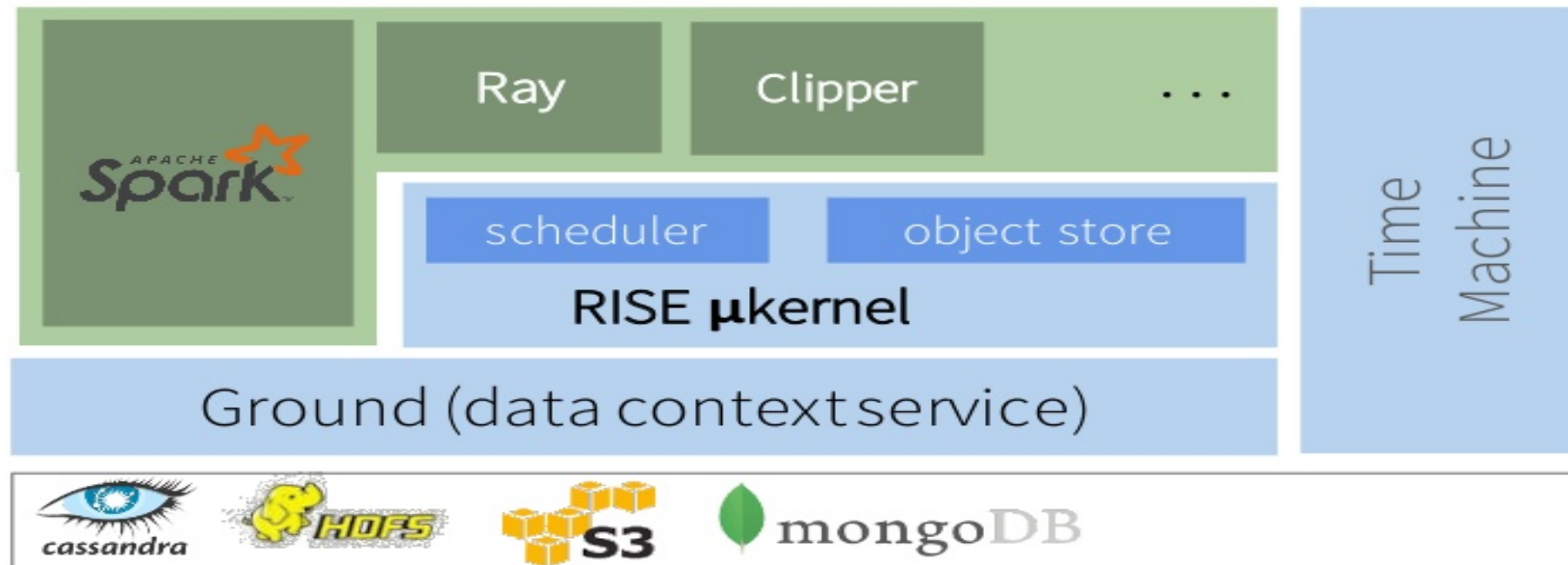
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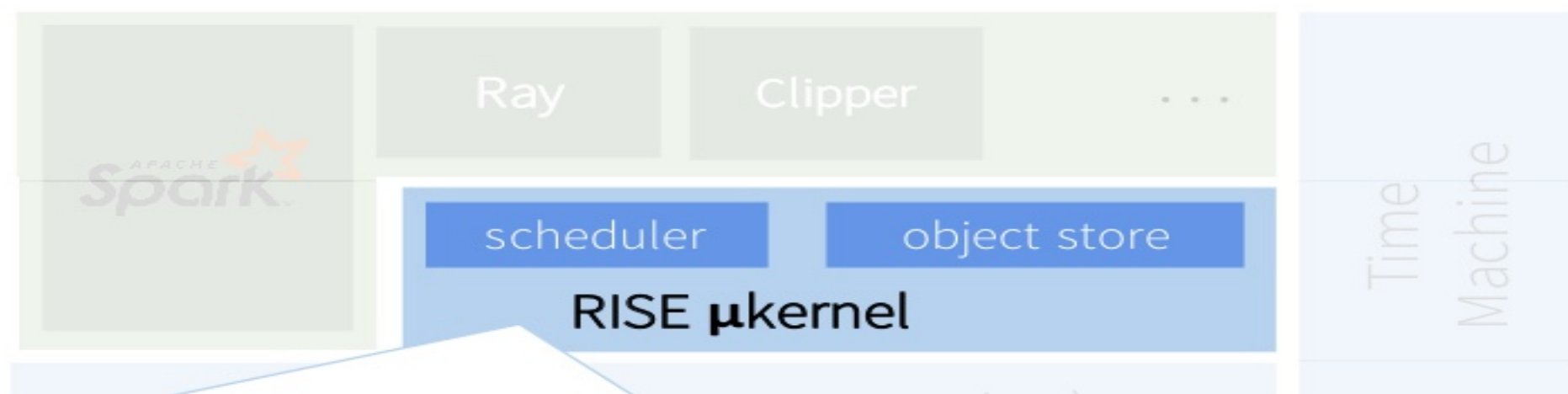
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Secure Real-time Decision Stack (SRDS)



Secure Real-time Decision Stack (SRDS)



Minimalist execution engine:

- Support both data flow and task-parallel execution models
- High-throughput, low-latency: ~ 1M tasks/sec @ ms latency

Secure Real-time Decision Stack (SRDS)

Central repository for models, APIs to capture the context in which data gets used and produced

Status: ongoing project with industry partners

Ground (data context service)



Secure Real-time Decision Stack (SRDS)

Replaying of apps at fine granularity

- Simplify development, debugging
- **Robustness**: replay against perturbed inputs
- **Explainability**: identify inputs causing decision
- **Security**: confirm vulnerabilities, test security patches, compliance auditing

Time
Machine



Secure Real-time Decision Stack (SRDS)



Dramatically simplify development of RISE applications

- Apache Spark: improve latency and security
- Clipper: model serving for Apache Spark, Scikit learn, etc
- Ray: framework for RL applications

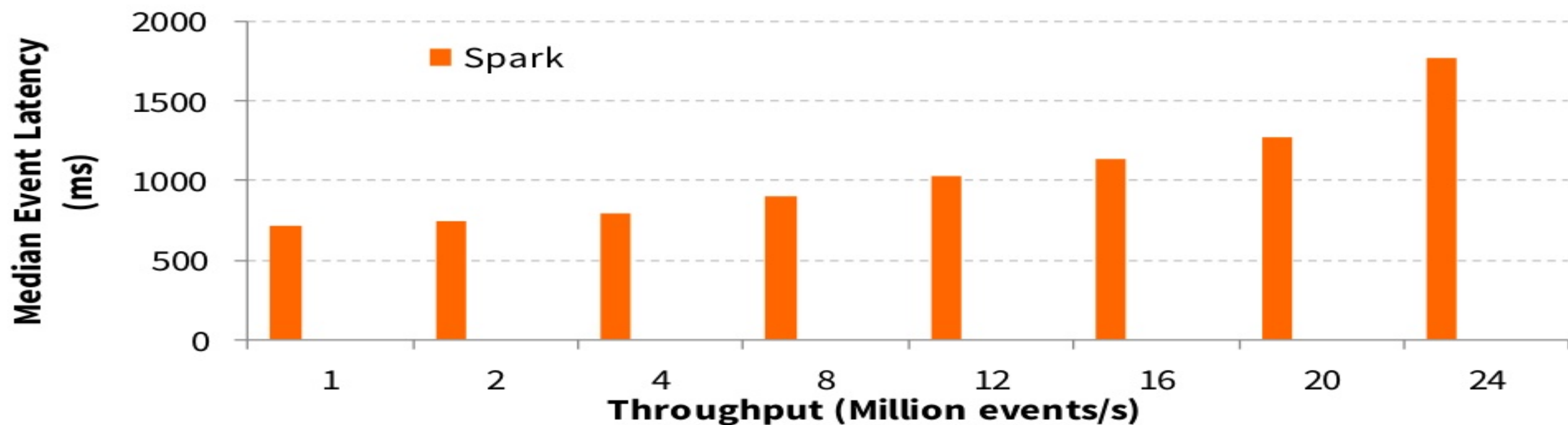


Improving Apache Spark

Drizzle

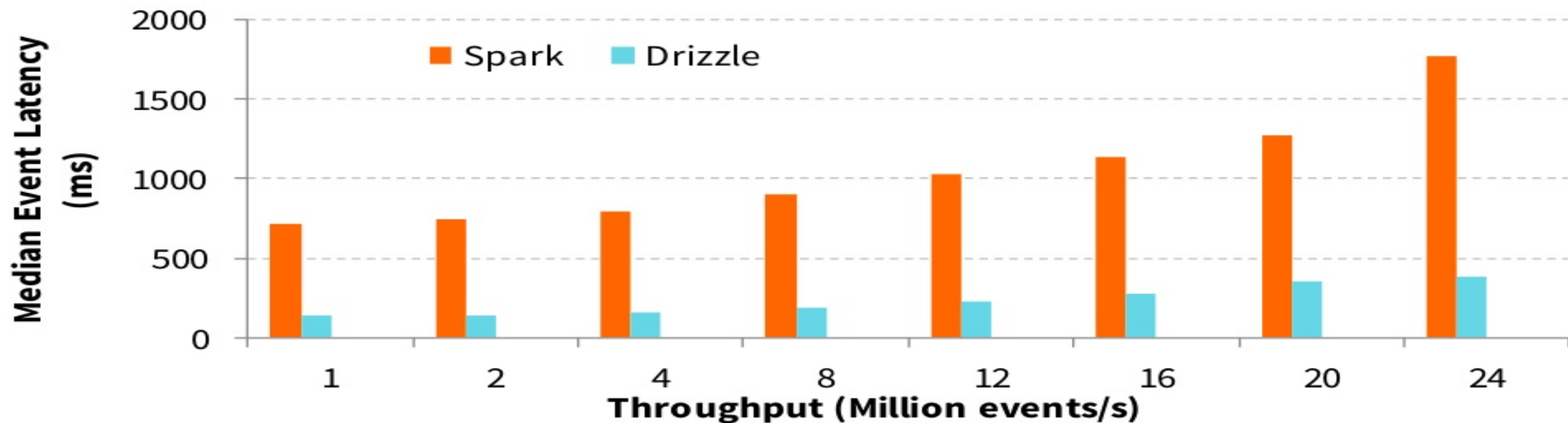
- Decrease latency of Structured Streaming and ML algorithms by ~10x
- Techniques: group scheduling, shared variables

Streaming Latency: YCSB benchmark



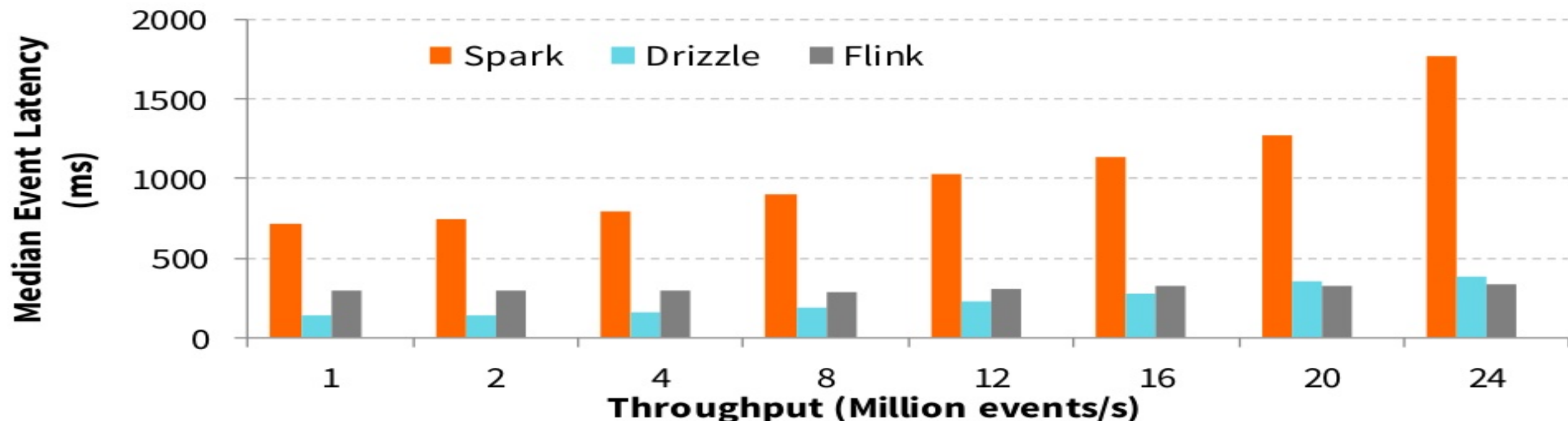
Drizzle-Opt: Reduce-by on mapper side

Streaming Latency: YCSB benchmark



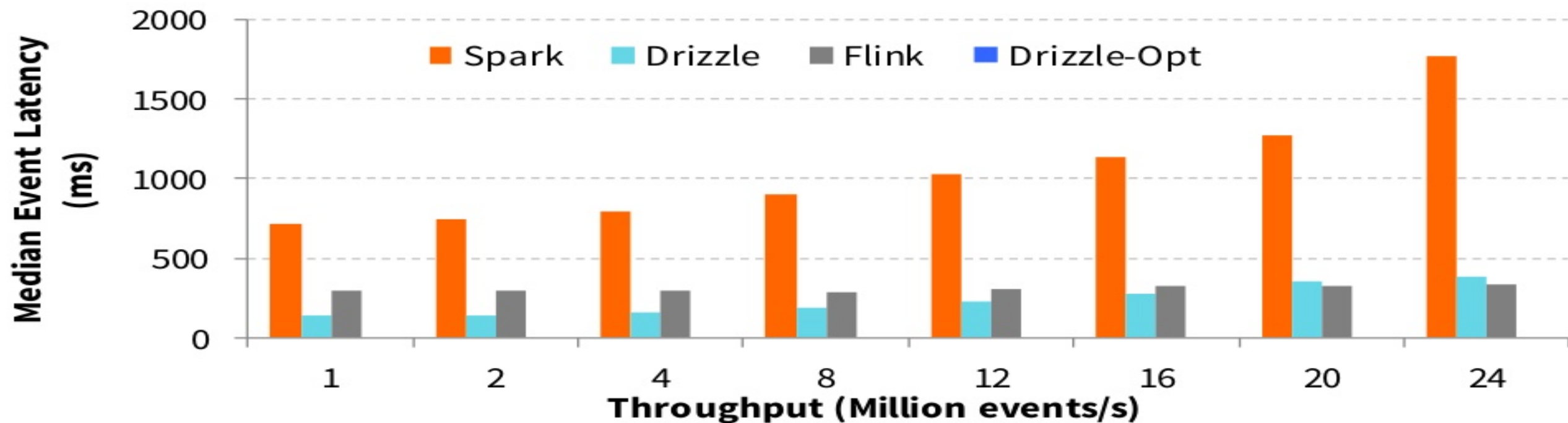
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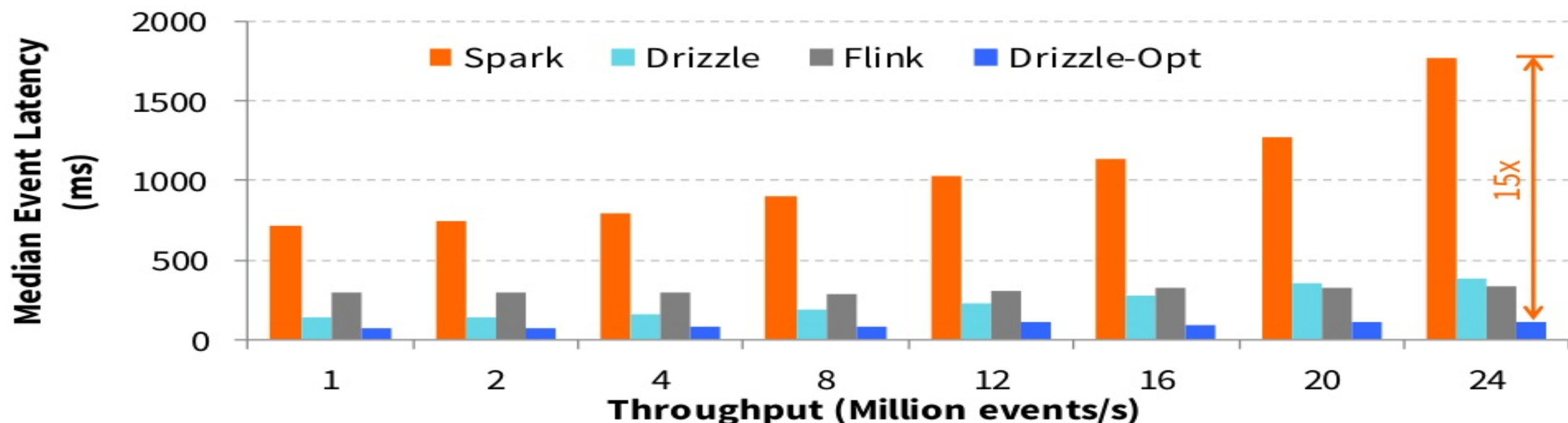
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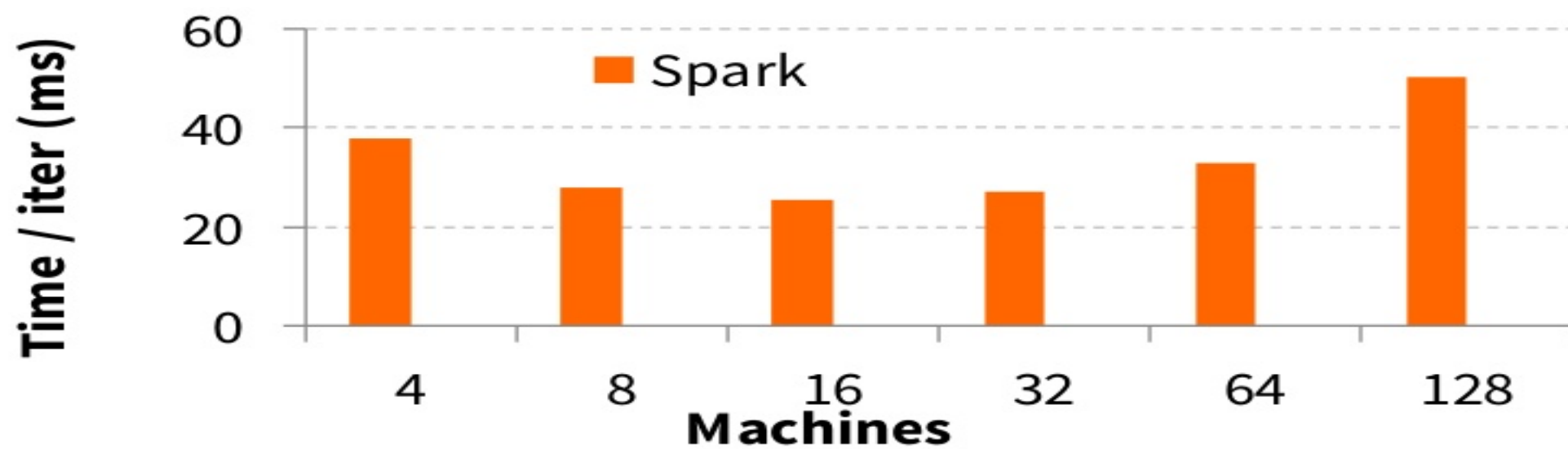
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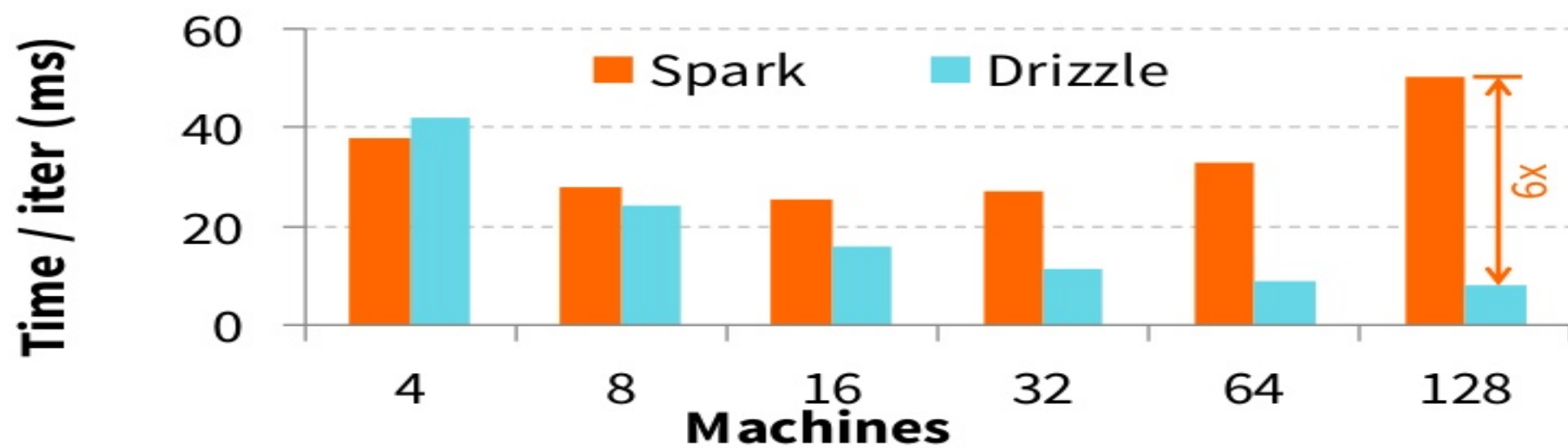


Drizzle-Opt: Reduce-by on mapper side

MLlib: SGD Performance



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Improving Apache Spark

Drizzle

- Decrease latency of Structured Streaming and ML algorithms by ~**10x**
- Techniques: group scheduling, shared variables
- **Some of these techniques will make their way to Apache Spark**

Opaque

- Full data encryption, authentication, and verification (Intel's SGX)
- Oblivious mode: hide data access pattern
- Support most SparkSQL functionality
- **See Wenting's talk later**

RISELab

Goal: Develop **open source** platforms, tools, and algorithms for intelligent real-time decisions on live-data

Already promising results

Expect much more over the next five years!

Thank you

