# ASYNC: A Cloud Engine with Asynchrony and History for Distributed Machine Learning

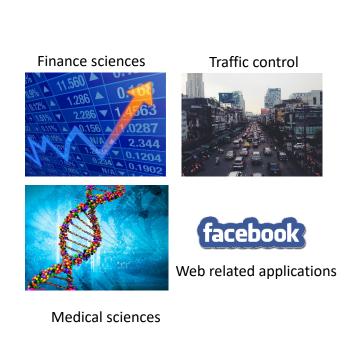
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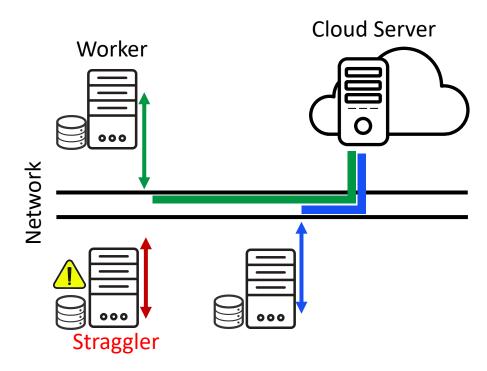
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## MACHINE LEARNING ON CLOUD

- ➤ Machine learning algorithms need to scale on cloud systems.
- Cloud systems are cheap but faulty.
- Nodes can be slow and stall the server.





#### **OPTIMIZATION IN MACHINE LEARNING**

- In many machine learning application, a model is obtained by applying an optimization algorithm on noisy data.
- Recent optimization algorithms must support asynchrony and history to mitigate the effect of slow workers and noise.
- Existing frameworks:
  - Create static dependency graphs to implement asynchrony<sup>1</sup>.
  - Store history using checkpointing or lineage-based methods<sup>2</sup>.

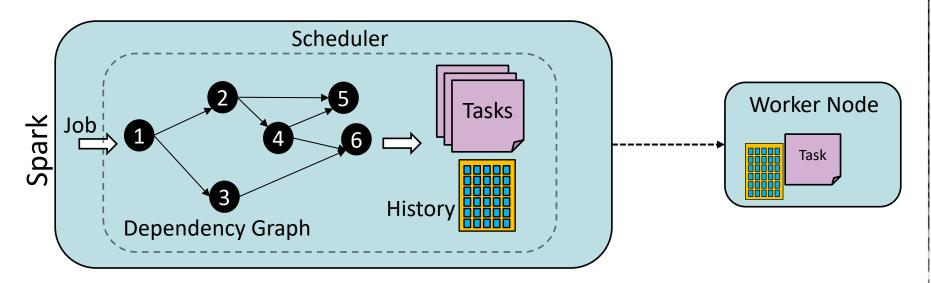
## **OPTIMIZATION IN MACHINE LEARNING**

- Problems with current tools:
  - Limited to asynchrony with fixed execution model throughout the algorithm.
  - Overhead of lineage and checkpointing is high.
- Our work:
  - Supports asynchrony by implementing dynamic dependency graphs.
  - Recovers history by partial model updates.
- We build our framework on top of Apache Spark<sup>1</sup>.

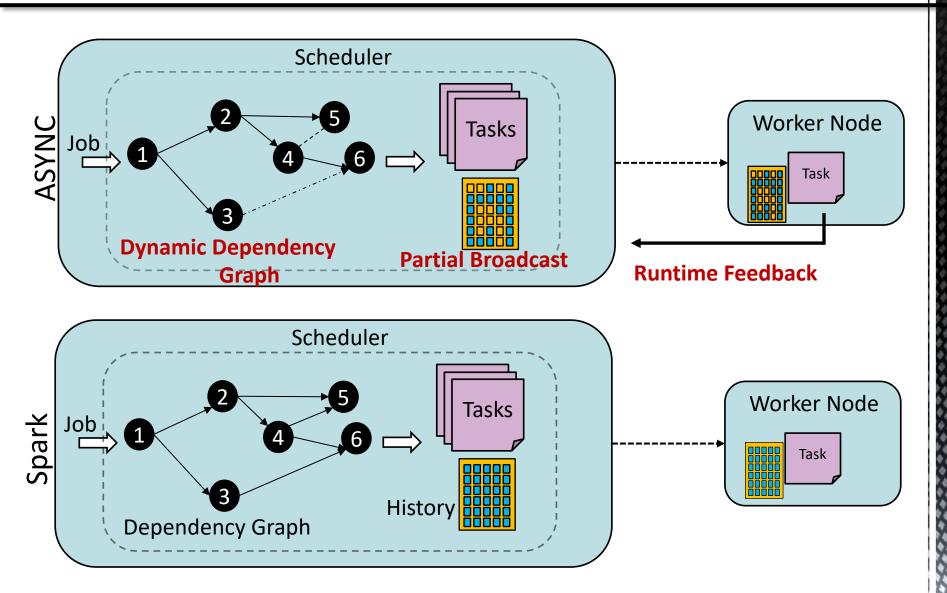
<sup>[1]</sup> M. Zaharia et al., "Resilient distributed datasets: A fault-tolerant abstraction for in-memory cluster computing," in Proceedings of the 9th USENIX conference on Networked Systems Design and Implementation. USENIX Association, 2012, pp. 2–2.

#### **ASYNC VS SPARK: INTERNALS**

- Spark uses a fixed dependency graph with periodic synchronization => relax the dependencies.
- It sends the bulky history with tasks => send partial history and recover the rest.



# **ASYNC VS SPARK: INTERNALS**



## **ASYNC VS SPARK: CODE**

- ASYNC uses a programming model written on top of Spark.
- In ASYNC, the user only adds the blue lines to convert a synchronous optimization method to an asynchronous variant.

```
Input: points, learning rate \alpha
Output: model \omega
for i = 1:N
                                                        for i = 1:N
     //broadcast model workers
     broadcast(\omega)
     //compute gradients
     \nabla f = \text{points.map}(...).\text{reduce}()
      //update model
      \omega = \omega - \alpha \nabla f
end
                                                        end
```

```
Input: points, learning rate \alpha
Output: model \omega
     //partial broadcast
     ASYNCbroadcast(\omega)
     //apply custom asynchrony
     points.ASYNCbarrier(...)
     \nabla f = \text{points.map}(...).\text{ASYNCreduce}()
     \omega = \omega - \alpha \nabla f
                     ASYNC
```

# **EXPERIMENTS**

- We implement two optimization algorithms in ASYNC and compare to their implementations in Spark:
  - Asynchronous stochastic gradient descent (ASGD)<sup>1</sup>: asynchronous without history.
  - ASAGA<sup>2</sup>: asynchronous with history.
- Datasets are epsilon and mnist8m with 400k and 8M samples each<sup>3</sup>.
- A cluster<sup>4</sup> of 32 workers with 8 straggler nodes is used.

<sup>[1]</sup> L. Bottou, "Stochastic gradient descent tricks," in Neural networks: Tricks of the trade. Springer, 2012, pp. 421–436.

<sup>[2]</sup> R. Leblond, F. Pedregosa, and S. Lacoste-Julien, "Asaga: asynchronous parallel saga," arXiv preprint arXiv:1606.04809, 2016.

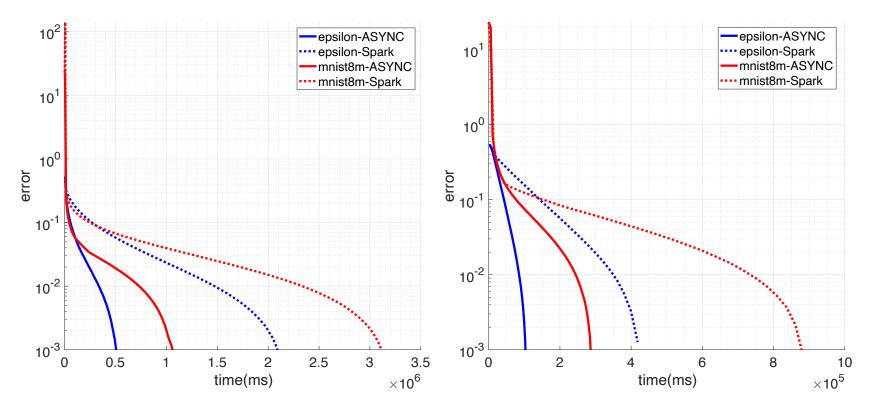
<sup>[3]</sup> C.-C. Chang and C.-J. Lin, "Libsvm: a library for support vector machines," ACM transactions on intelligent systems and technology (TIST), vol. 2, no. 3, p. 27, 2011.

<sup>[4]</sup> J. Towns et al., "Xsede: accelerating scientific discovery," Computing in Science & Engineering, vol. 16, no. 5, pp. 62–74, 2014.

# **RESULTS**



# ASAGA algorithm



ASYNC is 4 times faster than Spark.

#### **RESULTS**

- We developed ASYNC, a framework to support asynchrony and history for optimization algorithms in machine learning.
- ASYNC outperforms current implementations and achieves 4x speedup compared to Apache Spark.
- The link to ASYNC:
  - https://github.com/ASYNCframework/ASYNCframework
- The link to audio of the presentation:
  - https://bit.ly/3bh88F6
  - Email: sasoori@cs.toronto.edu