

# Mitigating the Impact of Tail Latency of Storage Systems on Scalable Deep Learning Applications

Hiroki Ohtsuji<sup>1</sup>, Erika Hayashi<sup>1</sup>, Naoto Fukumoto<sup>1</sup>, Eiji Yoshida<sup>1</sup>  
Takuya Okamoto<sup>2</sup>, Takeru Kuramoto<sup>3</sup>, Osamu Tatebe<sup>3</sup>

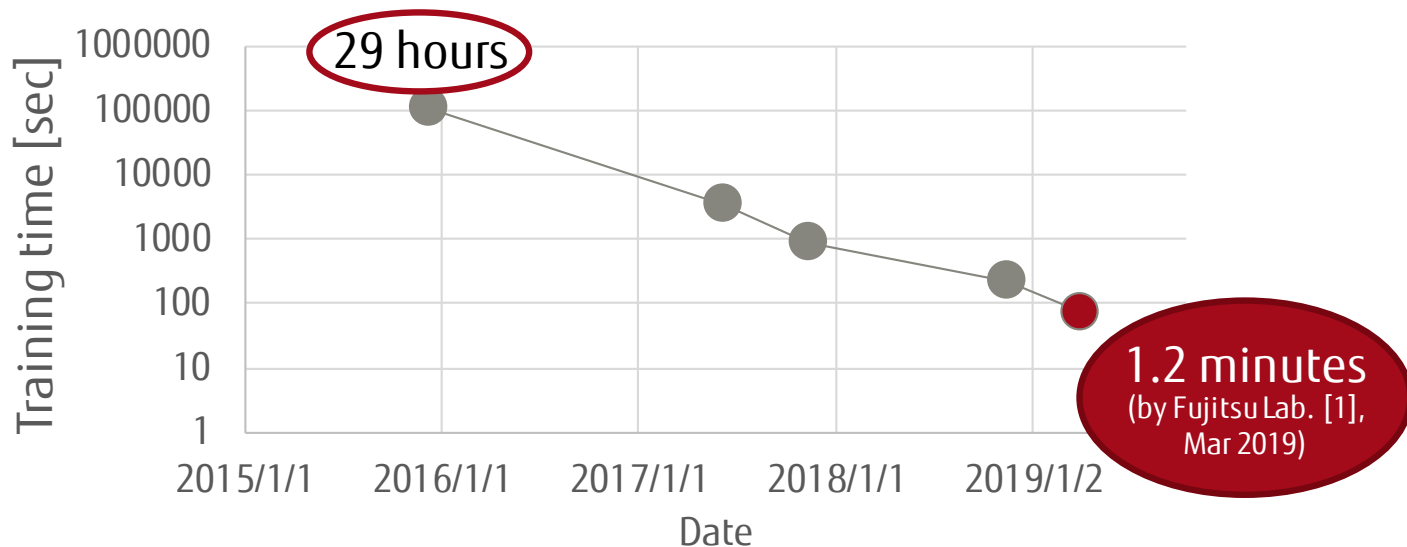
<sup>1</sup>Fujitsu Laboratories Ltd.

<sup>2</sup>Fujitsu Ltd.

<sup>3</sup>University of Tsukuba

# Massive-scale Fast Distributed Deep Learning

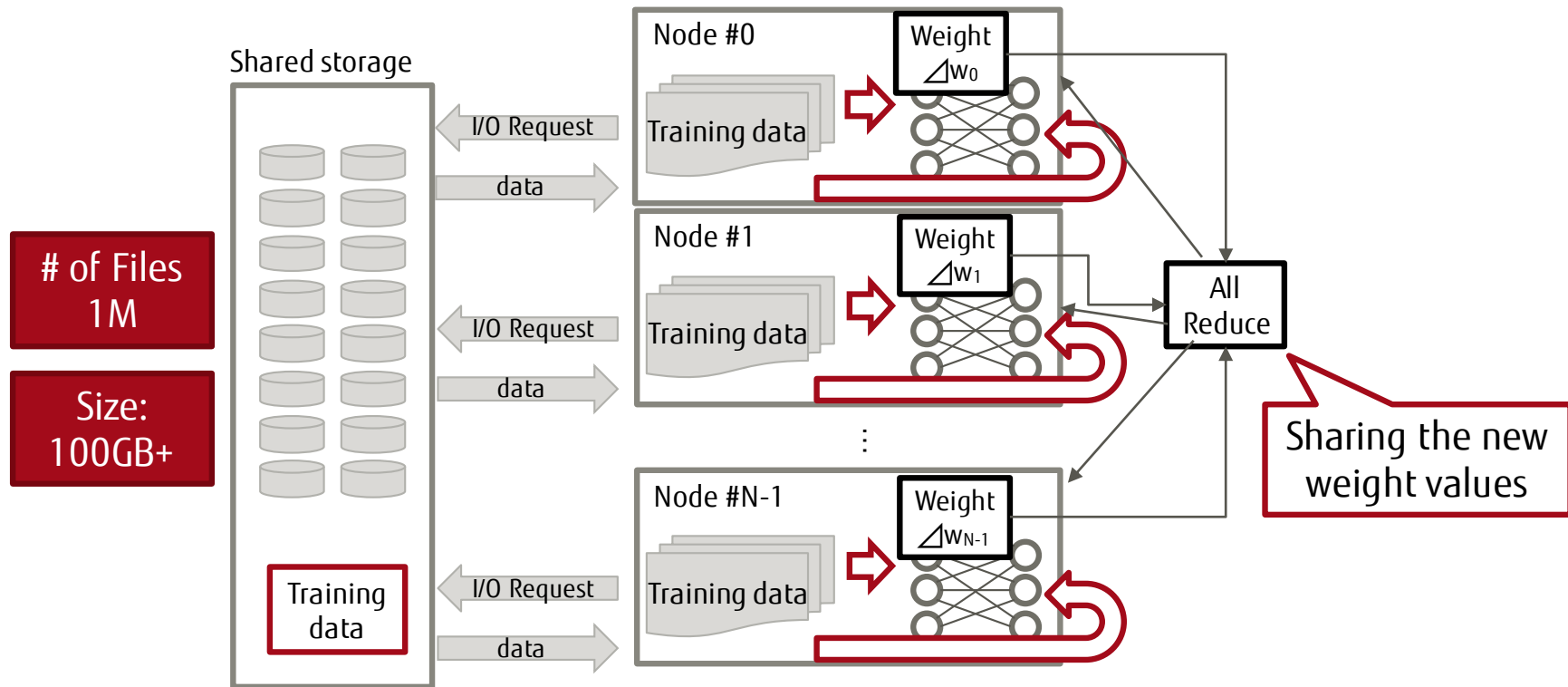
- ResNet-50/ImageNet (1000x1000 Images, 120 GB) training in **a minute**
  - All training processes read the data sets during the training
  - Can shared storage systems provide enough access performance to those workloads?



[1] Masafumi Yamazaki et al. Yet Another Accelerated SGD: ResNet-50 Training on ImageNet in 74.7 seconds. *CoRR*, abs/1903.12650, 2019.

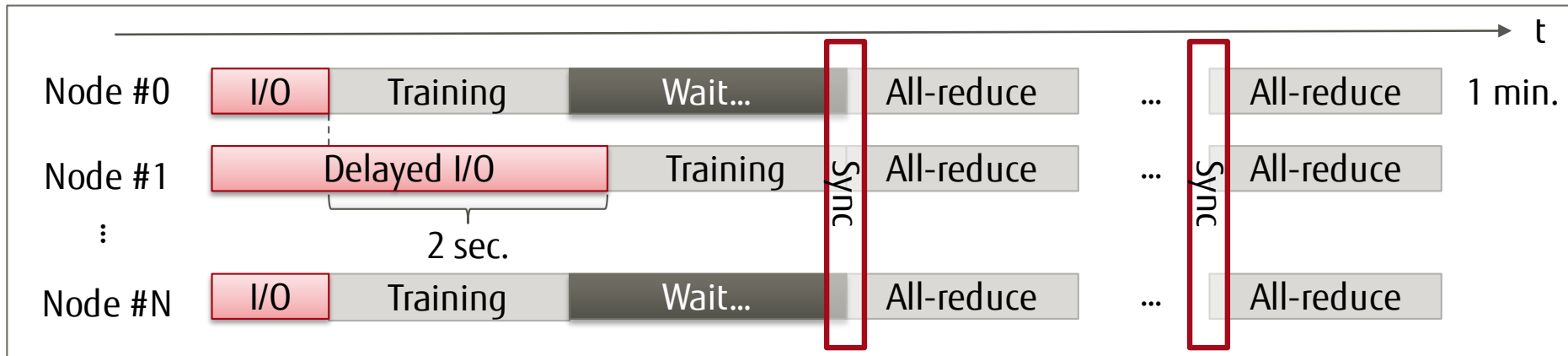
# Data parallel Distributed Deep Learning

- More than 1K processes read training data sets from a shared storage



# The Impact of Delayed I/O Requests

- A single delayed I/O can block the entire training process
  - A few seconds of tail latency is harmful due to the shortened training time
    - If the training takes **26 hours**, a 2 seconds delay does not matter.
    - However, a 2 seconds delay in a **1-minute** training process is **significant**.

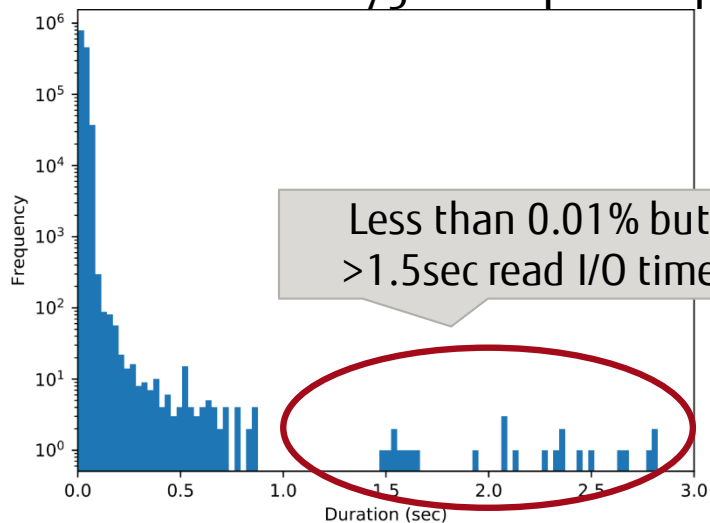


Synchronized processing makes the workload more sensitive to delayed I/Os.

# Observation: Tail Latency of Storage Systems

- Latency distribution when 16 processes read the ImageNet data set from the shared storage system

■ Measured on Cygnus Supercomputer at Univ. of Tsukuba

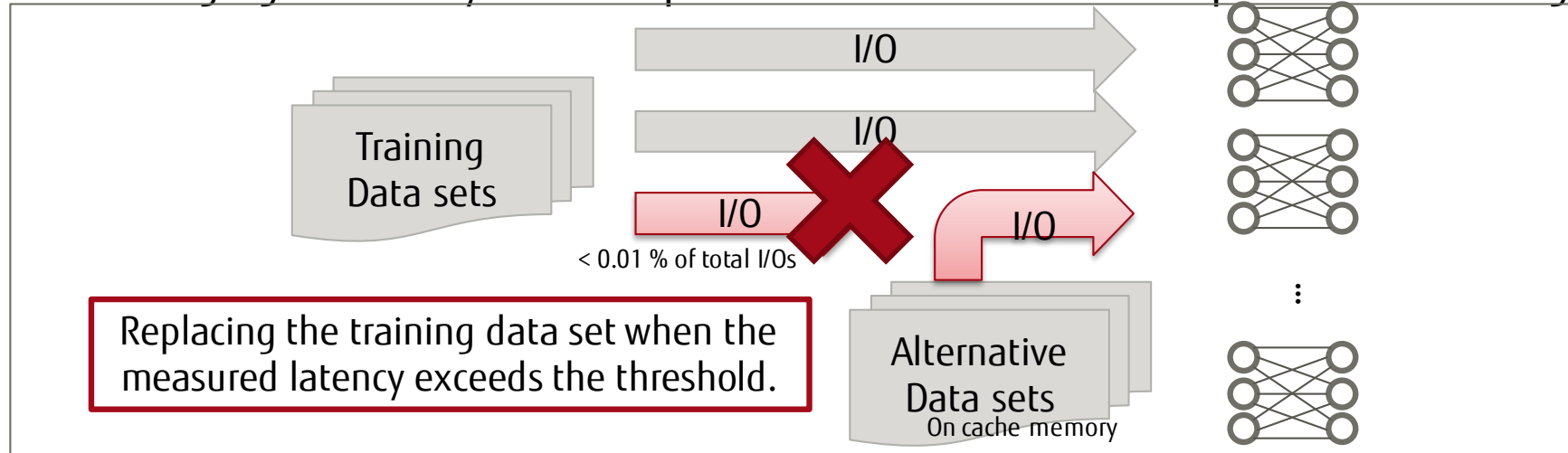


>99.99% of requests complete in **0.5 sec**  
<0.01% of requests take more than **1.5 sec**

There are very rare (<0.01%) long-delayed I/O requests on a shared storage system


# I/O Latency Aware Dynamic I/O Replacement

- Monitoring all read I/O operations and...
  - Discarding/replacing the delayed I/O with alternative data sets
  - Managing the history of data replacement to minimize the impact on the training.



Avoiding the impact of (rare) long-delayed I/O  
with the negligible drawback in precision of the trained model

- I/O tail latency has a significant impact on the performance of massive scale fast deep learning applications.
- Shared storage systems have very rare long tail latency, that affects to the performance of distributed deep learning workloads.
- We are developing a method to avoid the impact of delayed I/O requests on training processes.
  - Dynamic I/O Replacement
    - Replacing less than 0.01 % of total I/O operations.
    - Adaptive data replacement to minimize the drawbacks in the precision of the trained model.



**FUJITSU**

shaping tomorrow with you