







# Toward Portable Distributed Deep Learning

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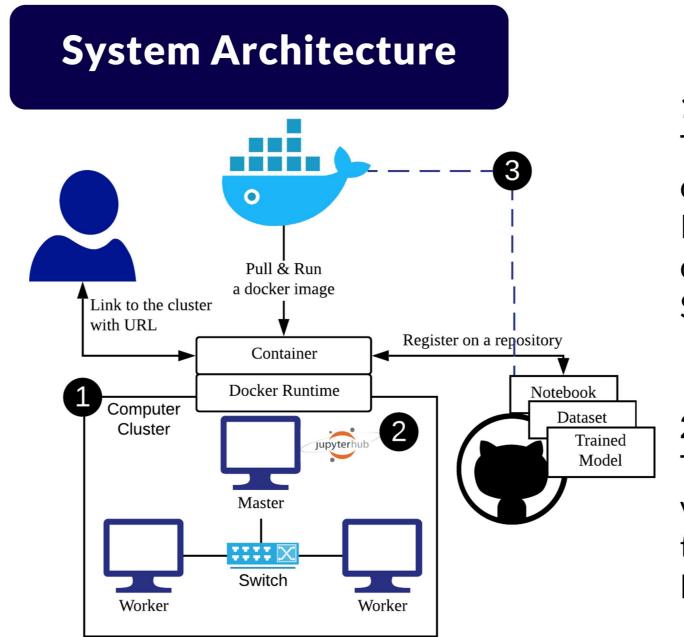
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# **Abstract**

Deep learning has become a key feature in several industries as a revolution in a field including image classification, speech recognition and machine translation. Deep learning algorithms normally consume a long processing time and a huge amount of computing resources; thus, they need special environments and devices to execute, thus they need special environments and devices to execute. Consequently, distributed training comes across to solve time consumption problem of deep learning by training dataset parallelly, but it still has a minimum requirement for devices to run an application in which is fixed to be executed on the specific device. This leads to the portable distributed deep learning idea which will allow other computers to access on a computer cluster anywhere and execute deep learning applications as distributed training on the server locally so that they can shorten their processing time due to data parallelization on the cluster.

#### **Portable Distributed Deep Learning** Bring your own distributed deep learning environment anywhere by leveraging container technologies **Container** repository e.g. DockerHub or NGC 1. Pull a container 3. Get JupyterHub URL 2. Launch a container image Container Container Notebook Singularity runtime Docker runtime Shared data Dataset repository trained local cluster **ABCI**



## 1 | Computer Cluster named 'mini-cluster'

mini-cluster is constructed with 3 computers which have 1 GPU, Nvidia Geforce RTX2080Ti, in each node, each node is connected with NETGEAR Gigabit Ethernet Switch for communicating data.

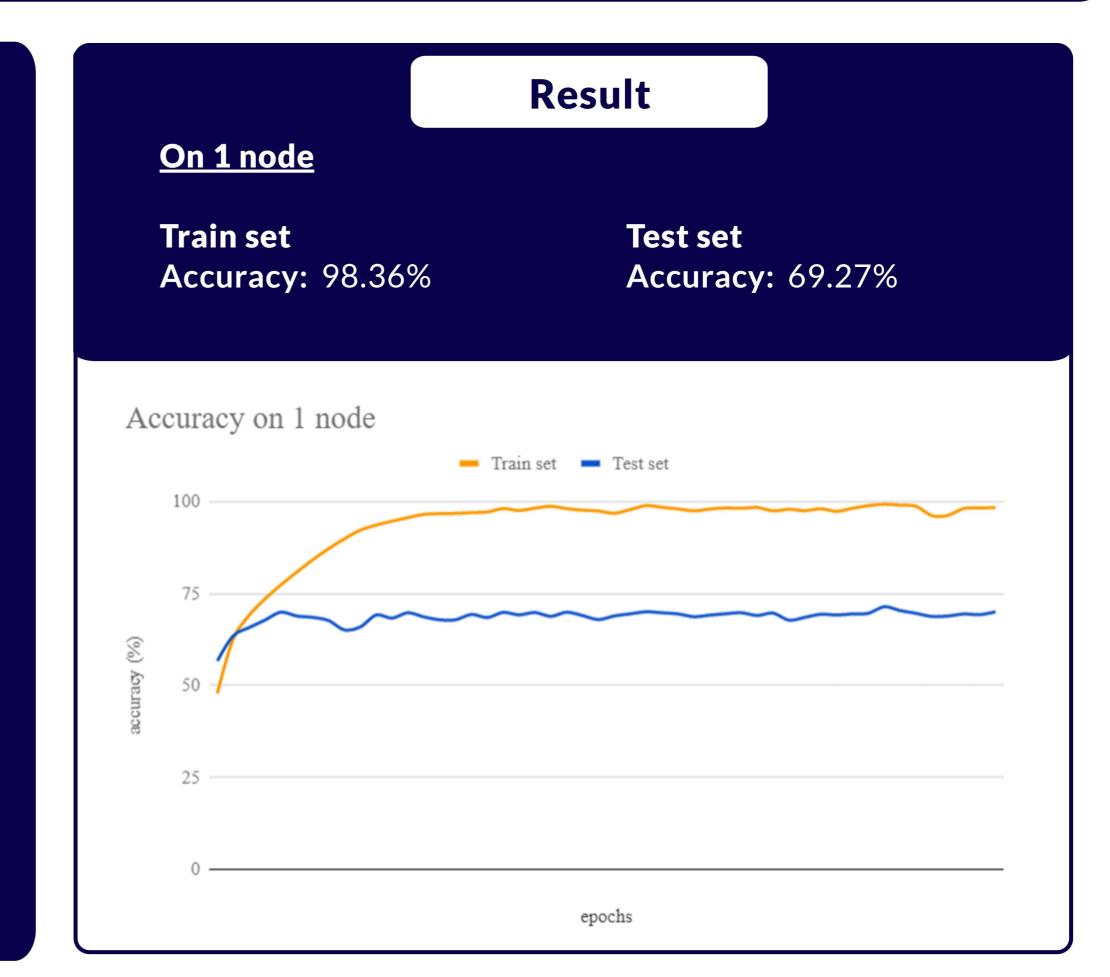
## 2 | JupyterHub

The JupyterHub package is installed with the virtual environment inside the mini-cluster; then, it will generate an URL for other users to link with the JupyterHub on the mini-cluster

## 3 | Docker with GitHub or Dockerhub repository

Docker is used to wrap up the JupyterHub package after configuration as a docker image and to register it on a repository on GitHub or Dockerhub for rebuild it other computer clusters by pulling and running the image.

age nvcr.io/nvidia/pytorch:19.08-py3	Dataset
	Docker Image
Dyforch I / UaU+fbaac4 I	Deep Learning Framework
nark https://github.com/ryousei/pytorch-cifar	Benchmark
CPU Intel Core i5-8400	CPU
nory DDR4-2666 32GB	Memory
SPU Nvidia GeForce RTX 2080Ti	GPU
OS Ubuntu 18.04	OS
ooch 50	Epoch
size 100	Batchsize
rate 0.01	Learning rate



## References

## Dataset from:

Darlow, L. and Crowley, E. (2018). CINIC-10 Is Not ImageNet or CIFAR-10. [online] Available at: https://arxiv.org/pdf/1810.03505v1.pdf

## **Background Information:**

Petuum, Inc.. (2018). Intro to Distributed Deep Learning Systems. [online] Available at: https://medium.com/@Petuum/intro-to-distributed-deep-learning-systems-a2e45c6b8e7 Project Jupyter team, (2016). JupyterHub — JupyterHub 1.0.0 documentation. [online] Available at: https://jupyterhub.readthedocs.io/en/stable/ Docker Documentation. (2019). Docker overview. [online] Available at: https://docs.docker.com/engine/docker-overview/

## Acknownledgement

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