



Example Kubeflow Operations and Tasks

NetApp Solutions

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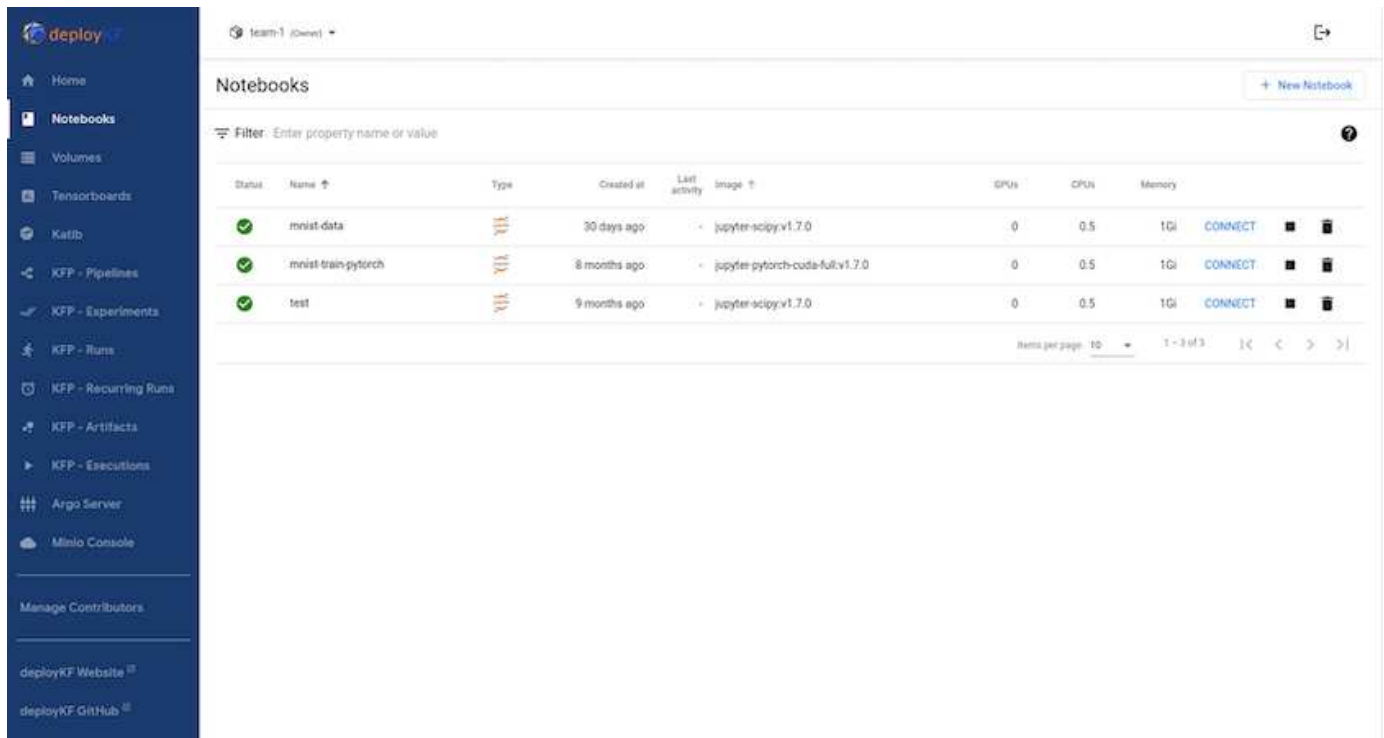
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Example Kubeflow Operations and Tasks

Provision a Jupyter Notebook Workspace for Data Scientist or Developer Use

Kubeflow is capable of rapidly provisioning new Jupyter Notebook servers to act as data scientist workspaces. For more information about Jupyter Notebooks within the Kubeflow context, see the [official Kubeflow documentation](#).



Use the NetApp DataOps Toolkit with Kubeflow

The [NetApp Data Science Toolkit for Kubernetes](#) can be used in conjunction with Kubeflow. Using the NetApp Data Science Toolkit with Kubeflow provides the following benefits:

- Data scientists can perform advanced NetApp data management operations, such as creating snapshots and clones, directly from within a Jupyter Notebook.
- Advanced NetApp data management operations, such as creating snapshots and clones, can be incorporated into automated workflows using the Kubeflow Pipelines framework.

Refer to the [Kubeflow Examples](#) section within the NetApp Data Science Toolkit GitHub repository for details on using the toolkit with Kubeflow.

Example Workflow - Train an Image Recognition Model Using Kubeflow and the NetApp DataOps Toolkit

This section describes the steps involved in training and deploying a Neural Network for Image Recognition using Kubeflow and the NetApp DataOps Toolkit. This is intended to serve as an example to show a training job that incorporates NetApp storage.

Prerequisites

Create a Dockerfile with the required configurations to use for the train and test steps within the Kubeflow pipeline.

Here is an example of a Dockerfile -

```
FROM pytorch/pytorch:latest
RUN pip install torchvision numpy scikit-learn matplotlib tensorboard
WORKDIR /app
COPY . /app
COPY train_mnist.py /app/train_mnist.py
CMD ["python", "train_mnist.py"]
```

Depending on your requirements, install all required libraries and packages needed to run the program. Before you train the Machine Learning model, it is assumed that you already have a working Kubeflow deployment.

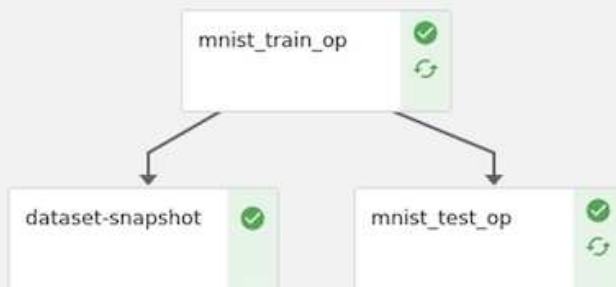
Train a Small NN on MNIST Data Using PyTorch and Kubeflow Pipelines

We use the example of a small Neural Network trained on MNIST data. The MNIST dataset consists of handwritten images of digits from 0-9. The images are 28x28 pixels in size. The dataset is divided into 60,000 train images and 10,000 validation images. The Neural Network used for this experiment is a 2-layer feedforward network. Training is executed using Kubeflow Pipelines. Refer to the documentation [here](#) for more information. Our Kubeflow pipeline incorporates the docker image from the Prerequisites section.

[←](#) ✔ mnist_pipeline 2024-04-03 15-57-35**Graph**

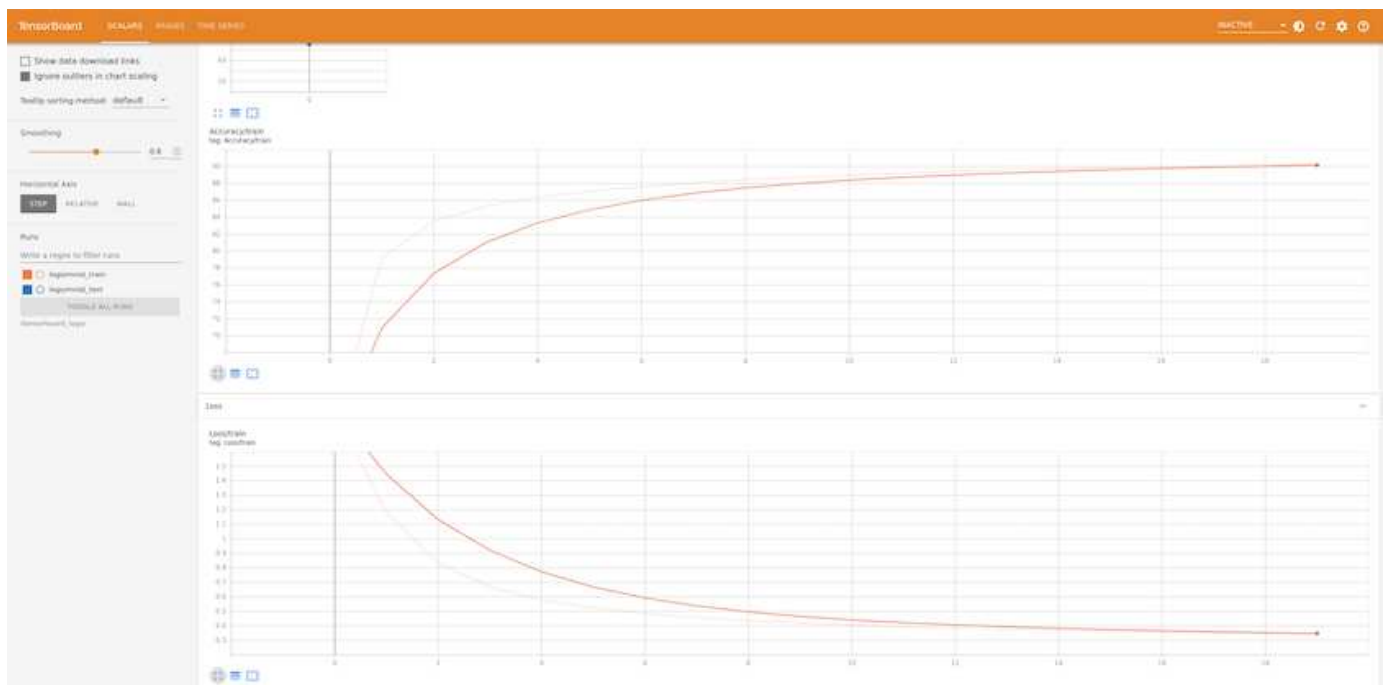
Run output

Config

 Simplify Graph

Visualize Results Using Tensorboard

Once the model is trained, we can visualize the results using Tensorboard. [Tensorboard](#) is available as a feature on the Kubeflow Dashboard. You can create a custom tensorboard for your job. An example below shows the plot of training accuracy vs. number of epochs and training loss vs. number of epochs.



Experiment with Hyperparameters Using Katib

Katib is a tool within KubeFlow that can be used to experiment with the model hyperparameters. To create an experiment, define a desired metric/goal first. This is usually the test accuracy. Once the metric is defined, choose hyperparameters that you would like to play around with (optimizer/learning_rate/number of layers). Katib does a hyperparameter sweep with the user-defined values to find the best combination of parameters that satisfy the desired metric. You can define these parameters in each section in the UI. Alternatively, you could define a **YAML** file with the necessary specifications. Below is an illustration of a Katib experiment -

The screenshot shows the Katib experiment configuration interface. The left sidebar contains navigation options: Home, Notebooks, Volumes, Tensorboards, Katib, KFP - Pipelines, KFP - Experiments, KFP - Runs, KFP - Recurring Runs, KFP - Artifacts, KFP - Executions, Argo Server, and Minio Console. The main content area displays the 'Experiment details' for an experiment named 'Validation-accuracy'. The objective is to maximize the validation accuracy to 0.9. The algorithm is set to 'grid'. Parameters include lr (0.01 to 0.03), num-layers (1 to 64), and optimizer (sgd, adam, ftrl). The metrics collector is set to 'File'.

The screenshot shows the Katib experiment overview interface. The left sidebar is the same as in the previous screenshot. The main content area displays the 'Experiment details' for an experiment named 'mnist-pytorch'. The status is 'Experiment is running'. The best trial performance is 'No optimal trial yet'. The user defined goal is 'Validation-accuracy > 0.9'. The overview table shows 3 running trials, 0 failed trials, and 0 succeeded trials.

OVERVIEW	TRIALS	DETAILS	YAML
Name	mnist-pytorch		
Status	⏱ Experiment is running		
Best trial	No optimal trial yet		
Best trial's params	No optimal trial yet		
Best trial performance			
User defined goal	Validation-accuracy > 0.9		
Running trials	3		
Failed trials	0		
Succeeded trials	0		

Use NetApp Snapshots to Save Data for Traceability

During the model training, we may want to save a snapshot of the training dataset for traceability. To do this, we can add a snapshot step to the pipeline as shown below. To create the snapshot, we can use the [NetApp](#)

DataOps Toolkit for Kubernetes.

```
@dsl.pipeline(
    name = 'MNIST Classification Pipeline',
    description = 'Train a simple NN for classification'
)
def mnist_pipeline():
    mnist_train_task = mnist_train_op()
    mnist_train_task.apply(
        kfp.onprem.mount_pvc('mnist-data', 'mnist-data-vol', '/mnt/data/')
    )

    mnist_test_task = mnist_test_op()
    mnist_test_task.apply(
        kfp.onprem.mount_pvc('mnist-data', 'mnist-data-vol', '/mnt/data/')
    )

    volume_snapshot_name = "mnist-pytorch-snapshot"
    dataset_snapshot = dsl.ContainerOp(
        name="dataset-snapshot",
        image="python:3.9",
        command=["/bin/bash", "-c"],
        arguments=[
            "python3 -m pip install netapp-dataops-k8s && \
            echo '' + volume_snapshot_name + '' > /volume_snapshot_name.txt && \
            netapp_dataops_k8s_cli.py create volume-snapshot --pvc-name=" + "mnist-data" + " --snapshot-name=" + str(volume_snapshot_name) + " --namespace={{workflow.namespace}}.",
            file_outputs={"volume_snapshot_name": "/volume_snapshot_name.txt"}
        ]
    )
    mnist_test_task.after(mnist_train_task)
    dataset_snapshot.after(mnist_train_task)
```

Refer to the [NetApp DataOps Toolkit example for KubeFlow](#) for more information.

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