

TO DO:

- Include screenshots
- Include detail/descriptions of what is happening
- Include other options not mentioned in tutorial (i.e. sampling types)
- Properly format document
- Proof read & spell check

MSE544-HyperDrive Experiment

HyperDrive

HyperDrive is a machine learning package found within Azure that aids in hyperparameter tuning/optimization. Hyperparameters are the parameters initialized before training that influence how the model trains and ultimately how the finished model performs. Examples of hyperparameters include: batch size, learning rate, number of layers in the neural network, the optimizer (e.g. Adam vs SGD), etc. Typically, the objective, when hyperparameter tuning, is to find the combination of hyperparameters that gives the best performing model. Azure has developed a package to make this discovery process much easier.

This tutorial will walk you through how to set up and run this process.

Repository Background

The framework presented in this work introduces the crystal graph convolution neural networks (CGCNN), which are designed to represent periodic crystal systems and predict material properties at DFT level accuracy and propose chemical insight. Read more about this study [here](#).

Dataset Introduction

A collection of 3,207 .cif crystal structures have been extracted from the "materials project" website and consolidated into an Azure data storage. [be more detailed](#)

Instructions

Part I: Set up the repository

1. Make a directory

```
mkdir MSE544-Hyperdrive
```

2. Move into the new directory

```
cd MSE544-Hyperdrive
```

3. Clone the repository we will be using

```
git clone https://github.com/txie-93/cgcnn.git
```

4. Move into the "cgcnn" directory

```
cd cgcnn
```

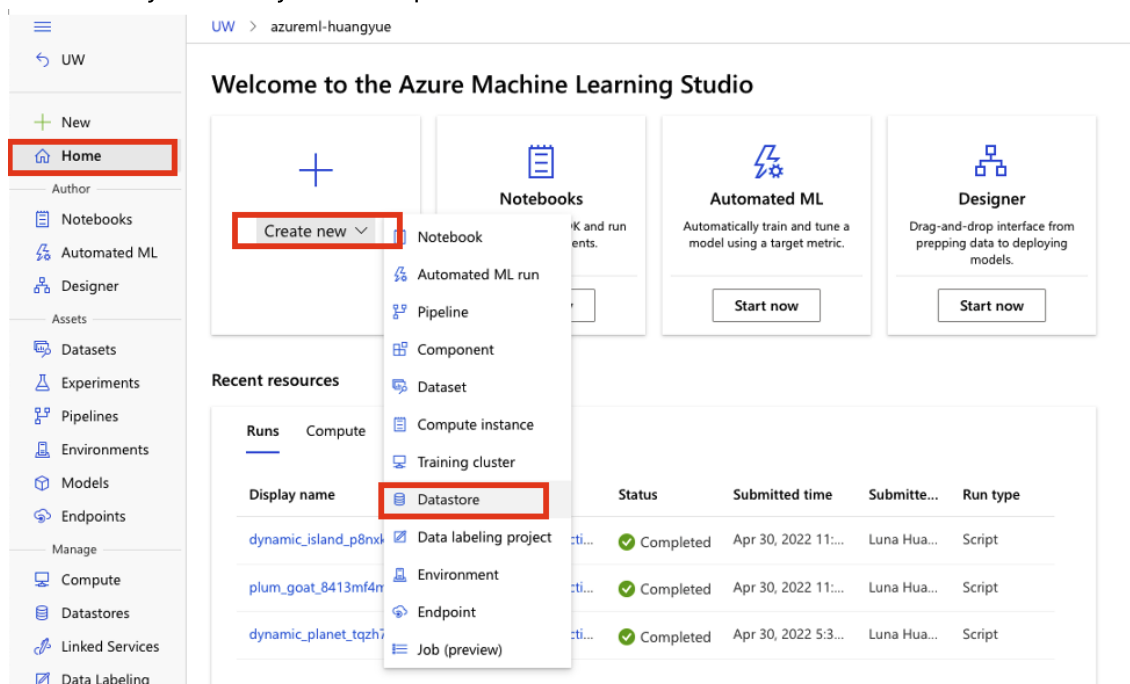
5. Starting at line 20 of the 'main.py' file, add the following two lines:

```
from azureml.core import Run
run = Run.get_context()
```

6. Download the .yml file from Canvas and place it in the "cgcnn" directory. The .yml (sometimes seen as .yaml) file is a special file typically used for configuring environments/settings for programs. Files with this extension are intended to be human-readable. FUN FACT: YAML initially stood for, *Yet Another Markdown Language*

Part II: Create an AML dataset linked to an Azure storage account

1. Create a data store in your ML workspace by click create/datastore from the homepage of ML studio, make sure you are in your workspace for this class.



2. Input all the information as shown in the screen shot below, and make sure you choose authentication type as SAS token (SAS aka Shared Access Signature), and copy paste SAS token `?sv=2020-08-04&ss=b&srt=co&sp=r&litfx&se=2022-07-02T02:55:20Z&st=2022-05-01T18:55:20Z&spr=https&sig=9P04kUW8p%2BsaX%2BJEkA%2FNuWX1f7TNp0iKr10S6dJAR`

M%3D, and then hit create. By creating a datastore, you link your workspace with an created storage account that already exists. In this way, multiple users can share the same data without having to copy the data intou your own workspace, therefore save the cost of data storage.

New datastore

Datastore name *

hyperdrivetutorial

Datastore type *

Azure Blob Storage

Account selection method

☐ From Azure subscription ☒ Enter manually

URL *

https://mse544storage.blob.core.windows.net/hyperdrivetutorialdata

Note: If you are attaching storage from different region than workspace region, it can result in...

Subscription ID

MSE544 Big Data and Informatics (ed4ef8f-68cc-4a10-9dd5-821829cbb45)

Resource group name of the storage resource

rg-amiclass-all

Save credentials with the datastore for data access

☐ No ☒ Yes

Authentication type *

SAS token

SAS token *

Use workspace managed identity for data preview and profiling in Azure Machine Learning studio

Create Cancel

3. Now let's create a dataset from datastore. In your ML studio home, click "Datasets"/"Create dataset"/"From datastore"

Microsoft Azure Machine Learning Studio

UW > azureml-huangyue > Datasets

Datasets

Registered datasets Dataset monitors (preview)

+ Create dataset

Refresh Unregister Edit columns Reset view

From local files

From datastore

From web files

From Open Datasets

	Version	Data source
molecule_images_yolov5	1	workspaceblobstore

4. Give a name to your dataset ``materials_hyperdrive_dataset`, select Dataset type as "File" and hit

Create dataset from datastore

☒ Basic info

☐ Datastore selection

☐ Confirm details

Basic info

Name *

materials_hyperdrive_dataset

Dataset type * ⓘ

File

Description

the dataset for cgcn project

Next

BackNext

5. In the prompt of Select or create a datastore, choose "hyperdrivetutorial" from the pull down menu (Note, since you have already link your datastore to the storage account, you should be able to select this existing one), and then choose the path to be ** and unclick "Skip data validation" hit next

Create dataset from datastore

☒ Basic info

☒ Datastore selection

☐ Confirm details

Datastore selection

Select or create a datastore * ⓘ

hyperdrivetutorial

*SQL type datastores are not supported for file type datasets

> Create new datastore

Path *

**

Browse

To include files in subfolders, append */** after the folder name like so: '(Folder)/**'. To filter by a file type, use *.extension, such as *.csv.

☐ Skip data validation ⓘ

6. Double check the information and hit Create

Create dataset from datastore

Basic info

Datastore selection

Confirm details

Basic info

Name

materials_hyperdrive_dataset

Dataset type

File

Description

the dataset for cgcnn project

Datastore selection

Datastore

hyperdrivetutorial

Path

**

Back

Create

7. Now if you go back to your ML workspace home and click datasets, you will be able to see the one you just created.

UW

New

Home

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UW > azureml-huangyue > Datasets

Datasets

Registered datasets

Dataset monitors (preview)

+ Create dataset

Refresh

Unregister

Edit columns

Reset view

Search

Showing 1-2 of 2 datasets

Name	Version	Data source	Created on	Modified on	Data type (...)	Properties	Created by
<input type="radio"/> materials_hyperdrive_dataset	1	hyperdrivetutorial	May 2, 2022 10:16 PM	May 2, 2022 10:16 PM	mltable	File	Luna Huang
<input type="radio"/> molecule_images_yolov5	1	workspaceblobstore	Apr 30, 2022 6:18 PM	Apr 30, 2022 6:18 PM	mltable	File	Luna Huang

8. Click the dataset, and click explore, you can see preview the files in your dataset.

UW

New

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Datastore

UW > azureml-huangyue > Datasets > materials_hyperdrive_dataset

materials_hyperdrive_dataset

Version 1 (latest)

Details

Consume

Explore

Models

New version

Refresh

Unregister

Preview

Number of files: 50 (sampled)

ID	Path	File Name	Modified Time	Created Time	File Size	File Format
1	/10000.cif	10000.cif	2022-05-02 15:48:42	2022-05-02 15:40:55	0.9395 KiB	.cif
2	/10003.cif	10003.cif	2022-05-02 15:48:42	2022-05-02 15:40:55	1.259 KiB	.cif
3	/10010.cif	10010.cif	2022-05-02 15:48:42	2022-05-02 15:40:55	0.9053 KiB	.cif
4	/1001011.cif	1001011.cif	2022-05-02 15:53:48	2022-05-02 15:53:48	1.436 KiB	.cif
5	/1001023.cif	1001023.cif	2022-05-02 15:53:48	2022-05-02 15:53:48	0.9326 KiB	.cif
6	/10015.cif	10015.cif	2022-05-02 15:48:42	2022-05-02 15:40:55	0.7383 KiB	.cif

Part III: Build the Notebook

1. Make a jupyter notebook called "hyperdrive_experiment"

- make sure this notebook is in the same directory as the "main.py" python script

2. Insert a cell with the following imports

```
from azureml.core import Workspace, Experiment, Environment,
ScriptRunConfig, Dataset, Run
import azureml
from azureml.train.hyperdrive import BayesianParameterSampling
from azureml.train.hyperdrive import normal, uniform, choice
from azureml.core.run import Run
from azureml.train.hyperdrive import HyperDriveConfig,
PrimaryMetricGoal
```

From the core tools package, we will import the standard classes for running jobs on Azure then we will import tools specific for hyperdrive to fine-tune our experiment.

3. Initialize a workspace in the next cell (be sure to enter the appropriate information)

```
subscription_id = <INSERT HERE>
resource_group  = <INSERT HERE>
workspace_name  = <INSERT HERE>
ws = Workspace(subscription_id, resource_group, workspace_name)
experiment = Experiment(workspace=ws, name='hyperdrive_experiment')
```

- *Workspace*: resource used for experimenting, training, and deploying machine learning models
- *Experiment*: defines the entry point for experiments in Azure. This is nothing more than a container that holds all of the runs you have submitted

4. Create a "dataset" variable that points to the data storage account holding the .cif files that we will use for training

```
dataset = Dataset.get_by_name(ws, name='materials_hyperdrive_dataset')
```

- *Dataset*: allows access to data in datastores (hosted on Azure) or from URLs that are publicly available

5. Set an environment variable using the .yaml file

```
cgcnn_env = Environment.from_conda_specification(name='cgcnn_env',
file_path='cgcnn_env.yaml')
```

- *Environment*: builds a reproducible python environment for the experiments to run in

6. Configure the base training session Here we are configuring our experiment, as we have done in previous tutorials.

```
config = ScriptRunConfig(source_directory='./',
                        script='main.py',
                        compute_target='<INSERT HERE>',
                        environment=cgcnn_env,
                        arguments=[
                            '--epochs', 5,
                            '--train-ratio', 0.6,
                            '--val-ratio', 0.2,
                            '--test-ratio', 0.2,
                            dataset.as_mount()]
                        )
```

- *ScriptRunConfig*: establishes the configuration information needed (python script, compute target, ...) to run the machine learning experiment
- *source_directory*: indicates the (working) directory our scripts can be found
- *script*: defines the python script we want to run
- *compute_target*: tells Azure where we want to run this experiment
- *environment*: initiates the predefined environment needed to successfully run this experiment
- *arguments*: allows us to define some constant parameters that the experiment should use (i.e. ratio of data allocated to the test, validation, and training set). Notice we also input our dataset here, which we have mounted

7. Define the parameters you are interested in sampling There are three different methods in which the hyperparameter space can be sampled:

- i. *Random sampling*: hyperparameters are randomly selected from the defined search space
- ii. *Grid sampling*: hyperparameters are selected such that all possible combinations are explored during experimentation (computationally expensive)
- iii. *Bayesian sampling*: hyperparameters are selected based on the outcomes of previous experiments; each subsequent run should be an improvement over the previous

In setting up our search space, we have the option of defining discrete or continuous hyperparameter spaces where the former is initiated by "choice" and the latter can be requested via "uniform" (amongst others)

```
param_sampling = BayesianParameterSampling( {
    "batch-size": choice(16, 32, 64),
    "learning-rate": uniform(0.05, 0.1),
    "optim": choice("SGD", "Adam")
})
```

(See reference I. for addition details)

8. Configure the hyperdrive experiment

```

hd_config = HyperDriveConfig(run_config=config,
                             hyperparameter_sampling=param_sampling,
                             primary_metric_name="MAE",

primary_metric_goal=PrimaryMetricGoal.MINIMIZE,
                             max_total_runs=8,
                             max_concurrent_runs=4)

```

- *HyperDriveConfig*: establishes the configuration information about the hyperparameters and relevant metrics needed to run the HyperDrive experiment
- *run_config*: instructions on how to set up the script runs (see step 6)
- *hyperparameter_sampling*: this is where the hyperparameter sampling space is specified, we have outlined the relevant hyperparameters and their respective spaces above in step 7
- *primary_metric_name*: define the metric of interest (in this case, we are interested in the mean absolute error, MAE)
- *primary_metric_goal*: decide how you want to evaluate your experiment (maximize or minimize the primary metric)
- *max_total_runs*: specify the number of runs you would like your experiment to complete (the default is 10080!)
- *max_concurrent_runs*: indicates the number of runs you would like to run concurrently, if a value is not specified, all runs will execute simultaneously

9. Finally, run the experiment and monitor the progress using the printed url

```

run = experiment.submit(hd_config)
aml_url = run.get_portal_url()
print(aml_url)

```

Part III: Running the Experiment and Navigating Azure

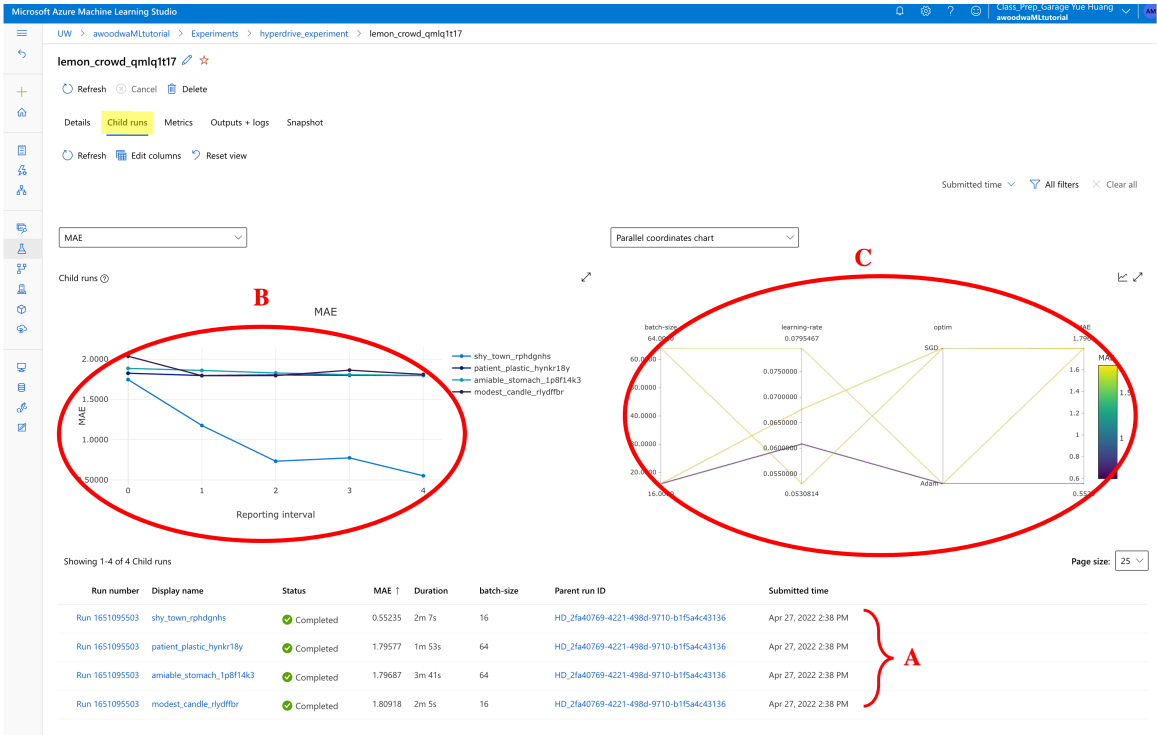
1. When you follow the url printed in step 9 of part II, you should find a page that looks something like

this:

A. Pathway to the experiment we are running

- B. Name of the current experiment - this is easily edited to something more meaningful by selecting the pencil symbol
- C. Tab showing the various runs that will be submitted during the experiment

2. Select the "child runs" tab to view the following page:



- A. Lists the subsequent runs within my experiment and provides relevant information such as: name of the run, status (pending, queued, complete), mean absolute error (MAE), duration of the run, batch size, time submitted. Notice the small arrow next to MAE, which indicates that I have sorted my runs based on the resulting MAE value.
- B. Visualization of the MAE for each run as they progressed
- C. Chart correlating the hyperparameters selected for each run and the calculated MAE
 - select the drop-down menu right above this plot to visualize the data in different dimensions

3. Select one of your child runs to further investigate by clicking on the display name

Microsoft Azure Machine Learning Studio

... > awoodwaMLtutorial > Experiments > hyperdrive_experiment > 220427_main.py_pc > patient_plastic_hynkr18y

patient_plastic_hynkr18y

Refresh Connect to compute Resubmit Cancel Delete

Details Metrics Images Child runs Outputs + logs Snapshot Explanations (preview) Fairness (preview) Monitoring (preview)

Properties

Status
✔ Completed

Created
Apr 27, 2022 2:38 PM

Started
Apr 27, 2022 2:49 PM

Duration
1m 52.86s

Compute duration
1m 52.86s

Run ID
HD_2fa40769-4221-498d-9710-b1f5a4c43136_1

Script name
main.py

Created by
Alexis Mills

Input datasets
Input name: input, Dataset: materials_project_3207_unzipped:1

Output datasets
None

Environment
cgcnv_env:Autosave_2022-04-27T16:43:26Z_5ce312c0

Arguments
--epochs 5 --train-ratio 0.6 --val-ratio 0.2 --test-ratio 0.2
DatasetConsumptionConfig:input --batch-size 64 --
learning-rate 0.05308136467463908 --optim SGD

Registered models
None

See all properties
Raw JSON

See YAML job definition
Job YAML

Tags

hyperparameters : {"batch-size": 64, "learning-rate": 0.05308136467463908, "optim": "SGD"}

Metrics

MAE
Min: 1.796, Max: 1.824, Last: 1.796

Description

Click edit icon to add a description

Compute

Target
GPU-awoodwa

Compute type
amlcompute

Instance count
1

We find a

lot of useful information here

- A. Here are the hyperparameters that were selected and used for this run, this is variable for every child run in this experiment
- B. An overview of the changing MAE is listed so we can easily see how it changed during the run
- C. Here are a list of the arguments that are given to our script, notice how are sampled hyperparameters from a. are used here an input

References

I. [Hyperparameter tuning models using Azure Machine Learning](#)