

ACES Graphcore Intelligence Processing Units (IPUs) Tutorial

07/14/2023

ACES Workshop

Chad Martin (Graphcore)

Zhenhua He (Texas A&M HPRC)



High Performance
Research Computing
DIVISION OF RESEARCH



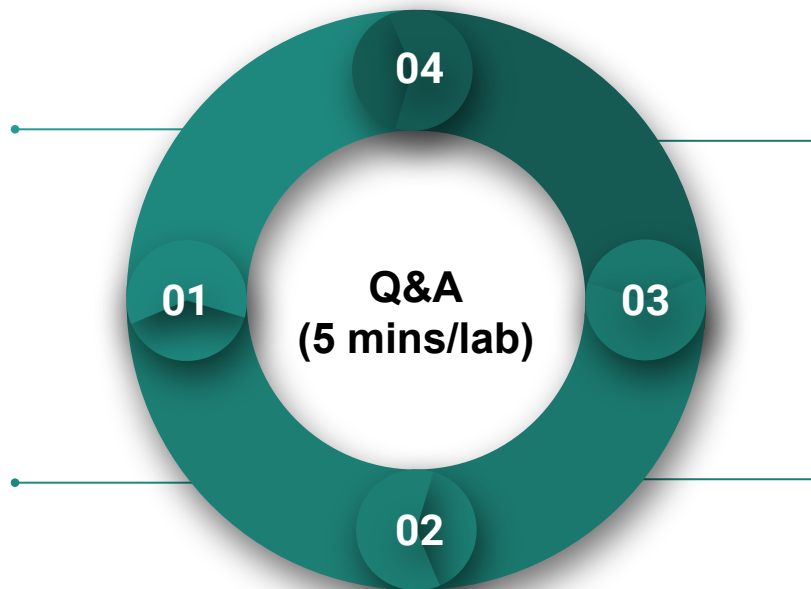
IPU Tutorial

Section I. Intro to IPU (30 mins)

We will introduce Graphcore IPU architecture, and the IPU systems on TAMU ACES platform.

Section II. Demo on ACES (30 mins)

We will demonstrate how to run models of different frameworks on ACES IPU system.



Section IV. Porting PyTorch code to IPU (30 minutes)

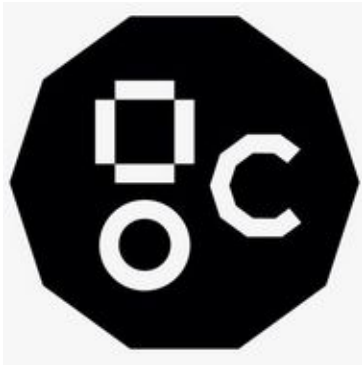
We will learn to port a PyTorch Fashion-MNIST classification model to run on IPU

Section III Porting TensorFlow code to IPU (30 minutes)

We will learn to port a Keras MNIST classification model to run on IPU

Structure of the IPU Tutorial.

Section I. ACES and IPU Systems Overview



ACES

Mission:

- Offer an accelerator testbed for numerical simulations and **AI/ML**
- Provide consulting, technical guidance, and training to researchers
- Collaborate on computational and data-enabled research.



Credit: towardsdatascience.com

Log into ACES Using the HPRC Portal

- HPRC webpage: <https://hprc.tamu.edu/>, Portal dropdown menu

ATM TEXAS A&M HIGH PERFORMANCE RESEARCH COMPUTING

Home User Services Resources Research Policies Events Training About **Portal**

Terra Portal
Grace Portal
FASTER Portal
FASTER Portal (ACCESS)
ACES Portal (ACCESS)

Quick Links

- New User Information
- Accounts
 - Apply for Accounts
 - Manage Accounts
- User Consulting
- Training
- Knowledge Base

TEXAS A&M UNIVERSITY TO ACQUIRE A

Accessing ACES via the ACES Portal (ACCESS)

Log-in using your ACCESS credentials.

ACCESS Powered By **CILogon**

Consent to Attribute Release

TAMU FASTER ACCESS OOD requests access to the following information. If you do not approve this request, do not proceed.

- Your CILogon user identifier
- Your name
- Your email address
- Your username and affiliation from your identity provider

Select an Identity Provider

ACCESS CI (XSEDE) + ?

☐ Remember this selection

Log On

By selecting "Log On", you agree to the [privacy policy](#).

For questions about this site, please see the [FAQs](#) or send email to help@cilogon.org.
Know your responsibilities using the CILogon Service. See [acknowledgments](#) for support for this site.

ACCESS

Login to CILogon

ACCESS Username

ACCESS Password

☐ Don't Remember Login

Login

CILogon

CILogon facilitates secure access to CyberInfrastructure (CI).

▲ If you had an XSEDE account, please enter your XSEDE username and password for ACCESS login

- > Register for an ACCESS Account
- > Forgot your password?
- > Need Help?

[Click Here for Assistance](#)

Select an Identity Provider

ACCESS CI (XSEDE) + ?

Select the Identity Provider appropriate for your account.

Shell access via the HPRC Portal

Access through (most) web browsers

-Top Banner Menu “Clusters” -> “Shell Access”



Training Materials

- From the ACES login node, ssh into the poplar2 (BOW Pod16) IPU system

```
ssh poplar2
```

- Change to your scratch directory:

```
cd $SCRATCH
```

- Copy the **example** materials to your scratch directory:

```
git clone https://github.com/graphcore/examples.git
```

- Copy the hands-on **exercise** materials to your scratch directory:

```
git clone https://github.com/happidencel/IPU-Training.git
```


Poplar SDK setup

```
source  
/opt/gc/poplar/poplar_sdk-ubuntu_20_04-3.3.0+1403-208993bbb  
7/poplar-ubuntu_20_04-3.3.0+7857-b67b751185/enable.sh
```

```
source  
/opt/gc/poplar/poplar_sdk-ubuntu_20_04-3.3.0+1403-208993bbb  
7/popart-ubuntu_20_04-3.3.0+7857-b67b751185/enable.sh
```

```
mkdir -p $SCRATCH/tmp  
export TF_POPLAR_FLAGS=--executable_cache_path=$SCRATCH/tmp  
export POPTORCH_CACHE_DIR=$SCRATCH/tmp
```

Run a TensorFlow (TF) model on IPU

TF Virtual Environment Setup

```
virtualenv -p python3 venv_tf2  
  
source venv_tf2/bin/activate  
  
python -m pip install -U pip  
  
python -m pip install  
/opt/gc/poplar/poplar_sdk-ubuntu_20_04-3.3.0+1403-208993bbb  
7/tensorflow-2.6.3+gc3.3.0+251582+08d96978c7f+intel_skylake  
512-cp38-cp38-linux_x86_64.whl
```

Run a TensorFlow model on IPU

```
cd examples/tutorials/tutorials/tensorflow2/keras/completed_demos/  
python completed_demo_ipu.py
```

- Deactivate the virtual environment after the model finishes running.

```
deactivate
```

Monitor the IPU usage with gc-monitor command

```
watch -n 2  
gc-monitor
```

Every 2.0s: gc-monitor				poplar2: Thu Jul 6 10:33:31 2023					
gc-monitor		Partition: p17 [active] has 16 reconfigurable IPU's							
IPU-M	Serial	IPU-M SW	Server version	ICU FW	Type	ID	IPU#	Routing	
10.5.5.1	0019.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	0	3	DNC	
10.5.5.1	0019.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	1	2	DNC	
10.5.5.1	0019.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	2	1	DNC	
10.5.5.1	0019.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	3	0	DNC	
10.5.5.2	0021.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	4	3	DNC	
10.5.5.2	0021.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	5	2	DNC	
10.5.5.2	0021.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	6	1	DNC	
10.5.5.2	0021.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	7	0	DNC	
10.5.5.3	0013.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	8	3	DNC	
10.5.5.3	0013.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	9	2	DNC	
10.5.5.3	0013.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	10	1	DNC	
10.5.5.3	0013.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	11	0	DNC	
10.5.5.4	0016.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	12	3	DNC	
10.5.5.4	0016.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	13	2	DNC	
10.5.5.4	0016.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	14	1	DNC	
10.5.5.4	0016.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	15	0	DNC	
Attached processes in partition p17				IPU			Board		
PID	Command	Time	User	ID	Clock	Temp	Temp	Power	
902631	python	50s	u.zh108696	0	1500MHz	23.5 C	21.9 C	90.3 W	

Run a PyTorch (PopTorch) model on IPU

PopTorch Virtual Environment Setup

```
cd $SCRATCH

virtualenv -p python3 poptorch_test

source poptorch_test/bin/activate

python -m pip install -U pip

python -m pip install
/opt/gc/poplar/poplar_sdk-ubuntu_20_04-3.3.0+1403-208993bbb
7/poptorch-3.3.0+113432_960e9c294b_ubuntu_20_04-cp38-cp38-l
inux_x86_64.whl
```

Run a PopTorch model on IPU

```
cd $SCRATCH/examples/tutorials/simple_applications/pytorch/mnist/  
  
pip install -r requirements.txt  
  
python mnist_poptorch.py
```

- Deactivate the virtual environment after the model finishes running.

```
deactivate
```


Monitor the IPU usage with gc-monitor command

```
watch -n 2  
gc-monitor
```

```
Every 2.0s: gc-monitor
```

poplar2: Thu Jul 6 10:55:55 2023

gc-monitor Partition: p17 [active] has 16 reconfigurable IPU's									
IPU-M	Serial	IPU-M SW	Server version	ICU FW	Type	ID	IPU#	Routing	
10.5.5.1	0019.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	0	3	DNC	
10.5.5.1	0019.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	1	2	DNC	
10.5.5.1	0019.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	2	1	DNC	
10.5.5.1	0019.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	3	0	DNC	
10.5.5.2	0021.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	4	3	DNC	
10.5.5.2	0021.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	5	2	DNC	
10.5.5.2	0021.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	6	1	DNC	
10.5.5.2	0021.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	7	0	DNC	
10.5.5.3	0013.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	8	3	DNC	
10.5.5.3	0013.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	9	2	DNC	
10.5.5.3	0013.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	10	1	DNC	
10.5.5.3	0013.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	11	0	DNC	
10.5.5.4	0016.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	12	3	DNC	
10.5.5.4	0016.0002.8222521	2.6.0	1.11.0	2.5.9	M2000	13	2	DNC	
10.5.5.4	0016.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	14	1	DNC	
10.5.5.4	0016.0001.8222521	2.6.0	1.11.0	2.5.9	M2000	15	0	DNC	
Attached processes in partition p17					IPU		Board		
PID	Command	Time	User	ID	Clock	Temp	Temp	Power	
907530	python	17s	u.zh108696	0	1500MHz	23.5 C	22.0 C	90.8 W	

Hands-On Session 1

- Please access ACES and poplar2 now.
- Copy the tutorial materials to your scratch directory.
- Run the TensorFlow and PyTorch (PopTorch) example models on IPU

Section III. Porting TensorFlow Code to IPU



TensorFlow



Keras

1. Import the TensorFlow IPU module

Add the following import statement to the beginning of your script:

```
from tensorflow.python import ipu
```

2. Preparing the dataset

- Make sure the sizes of the datasets are divisible by the batch size

```
def make_divisible(number, divisor):  
    return number - number % divisor
```

- Adjust dataset lengths

```
(x_train, y_train), (x_test, y_test) = load_data()  
train_data_len = x_train.shape[0]  
train_data_len = make_divisible(train_data_len, batch_size)  
x_train, y_train = x_train[:train_data_len], y_train[:train_data_len]  
test_data_len = x_test.shape[0]  
test_data_len = make_divisible(test_data_len, batch_size)  
x_test, y_test = x_test[:test_data_len], y_test[:test_data_len]
```

3. Add IPU configuration

To use the IPU, you must create an IPU session configuration:

```
ipu_config = ipu.config.IPUConfig()  
ipu_config.auto_select_ipus = 1  
ipu_config.configure_ipu_system()
```

A full list of configuration options is available in the [API documentation](#).

4. Specify IPU strategy

```
strategy = ipu.ipu_strategy.IPUStrategy()
```

The `tf.distribute.Strategy` is an API to distribute training and inference across multiple devices. `IPUStrategy` is a subclass which targets a system with one or more IPUs attached.

5. Wrap the model within the IPU strategy scope

- Creating variables and Keras models within the scope of the `IPUStrategy` object will ensure that they are placed on the IPU.
- To do this, we create a `strategy.scope()` context manager and move all the model code inside it.

Hands-on Session 2

- Activate the TF virtual environment

cd \$SCRATCH

source venv_tf2/bin/activate

- Change directory to Keras

cd IPU-Training/Keras

- Complete the **#Todos** in the mnist-ipu-todo.py file.

- Run it in the **venv_tf2** virtual environment.

python mnist-ipu-todo.py

- After finishing the job, you can deactivate the virtual environment

deactivate

Section IV. Porting PyTorch Code to IPU



PopTorch

- PopTorch is a set of extensions for PyTorch released by Graphcore to enable PyTorch models to run on Graphcore's IPU hardware.
- PopTorch will use PopART to parallelise the model over the given number of IPUs. Additional parallelism can be expressed via a replication factor, which enables you to data-parallelise the model over more IPUs.

Training a model on IPU

- Import the packages

```
import torch
import poptorch
import torchvision
import torch.nn as nn
import matplotlib.pyplot as plt
from tqdm import tqdm
from sklearn.metrics import accuracy_score
```

Load the data

PopTorch offers an extension of `torch.utils.data.DataLoader` class with its `poptorch.DataLoader` class, specialized for the way the underlying PopART framework handles batching of data.

Build the model

```
class ClassificationModel(nn.Module):
    def __init__(self):
        super().__init__()
        self.conv1 = nn.Conv2d(1, 5, 3)
        self.pool = nn.MaxPool2d(2, 2)
        self.conv2 = nn.Conv2d(5, 12, 5)
        self.norm = nn.GroupNorm(3, 12)
        self.fc1 = nn.Linear(972, 100)
        self.relu = nn.ReLU()
        self.fc2 = nn.Linear(100, 10)
        self.log_softmax = nn.LogSoftmax(dim=1)
        self.loss = nn.NLLLoss()
```

```
def forward(self, x, labels=None):
    x = self.pool(self.relu(self.conv1(x)))
    x = self.norm(self.relu(self.conv2(x)))
    x = torch.flatten(x, start_dim=1)
    x = self.relu(self.fc1(x))
    x = self.log_softmax(self.fc2(x))
    # The model is responsible for the
calculation of the loss when using an IPU. We do
it this way:
    if self.training:
        return x, self.loss(x, labels)
    return x
```

```
model = ClassificationModel()
model.train()
```

Prepare training for IPU

The compilation and execution on the IPU can be controlled using `poptorch.Options`. These options are used by PopTorch's wrappers such as `poptorch.DataLoader` and `poptorch.trainingModel`.

```
opts = poptorch.Options()
train_dataloader = poptorch.DataLoader(
    opts, train_dataset, batch_size=16, shuffle=True, num_workers=20
)
```

Train the model

```
optimizer = poptorch.optim.SGD(model.parameters(), lr=0.001, momentum=0.9)

poptorch_model = poptorch.trainingModel(model, options=opts,
optimizer=optimizer)

epochs = 30
for epoch in tqdm(range(epochs), desc="epochs"):
    total_loss = 0.0
    for data, labels in tqdm(train_dataloader, desc="batches", leave=False):
        output, loss = poptorch_model(data, labels)
        total_loss += loss

poptorch_model.detachFromDevice()

torch.save(model.state_dict(), "classifier.pth")
```


Evaluate the model

```
model = model.eval()

poptorch_model_inf = poptorch.inferenceModel(model, options=opts)

test_dataloader = poptorch.DataLoader(opts, test_dataset,
batch_size=32, num_workers=10)

predictions, labels = [], []
for data, label in test_dataloader:
    predictions += poptorch_model_inf(data).data.max(dim=1).indices
    labels += label

poptorch_model_inf.detachFromDevice()

print(f"Eval accuracy: {100 * accuracy_score(labels,
predictions):.2f}%")
```

Hands-on Session 3

- Activate the TF virtual environment

```
cd $SCRATCH
```

```
source poptorch_test/bin/activate
```

- Change directory to PyTorch

```
cd IPU-Training/PyTorch
```

- Complete the **#Todos** in the fashion-mnist-pytorch-ipu-todo.py file.
- Run it in the **poptorch_test** virtual environment.

```
pip install scikit-learn
```

```
python fashion-mnist-pytorch-ipu-todo.py
```

- After finishing the job, you can deactivate the virtual environment

```
deactivate
```

References

- <https://www.graphcore.ai/>
- <https://github.com/graphcore/examples/tree/v3.2.0/tutorials/tutorials/tensorflow2/keras>
- <https://github.com/graphcore/examples/tree/v3.2.0/tutorials/tutorials/pytorch/basics>
- https://hprc.tamu.edu/wiki/Main_Page



**HIGH PERFORMANCE
RESEARCH COMPUTING**
TEXAS A&M UNIVERSITY

<https://hprc.tamu.edu>

HPRC Helpdesk:

help@hprc.tamu.edu

Phone: 979-845-0219

Help us help you. Please include details in your request for support, such as, Cluster (ACES, Faster, Grace, Terra, ViDaL), Job information (Job id(s), Location of your jobfile, input/output files, Application, Module(s) loaded, Error messages, etc), and Steps you have taken, so we can reproduce the problem.