ACES Graphcore Intelligence Processing Units (IPUs) Tutorial

07/14/2023
ACES Workshop
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Zhenhua He (Texas A&M HPRC)
Section I. Intro to IPUs
We will introduce Graphcore IPU architecture, and the IPU systems on TAMU ACES platform.

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

Section III. Porting TensorFlow code to IPU
We will learn to port a Keras MNIST classification model to run on IPU.

Section IV. Porting PyTorch code to IPU
We will learn to port a PyTorch Fashion-MNIST classification model to run on IPU.

Q&A (10 mins/sec)

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/](https://hprc.tamu.edu/), Portal dropdown menu
Accessing ACES via the ACES Portal (ACCESS)

Log-in using your ACCESS credentials.

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+1403-208993bbb7/poplar-ubuntu_20_04-3.3.0+7857-b67b751185/enable.sh

source /opt/gc/poplar/poplar_sdk-ubuntu_20_04-3.3.0+1403-208993bbb7/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+1403-208993bbb7/tensorflow-2.6.3+gc3.3.0+251582+08d96978c7f+intel_skylake512-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
```

<table>
<thead>
<tr>
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Attached processes in partition p17

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Run a PyTorch (PopTorch) model on IPU
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+1403-208993bbb7/poptorch-3.3.0+113432_960e9c294bUbuntu_20_04-cp38-cp38-linux_x86_64.whl
Run a PopTorch model on IPU

```bash
cd $SCRATCH/examples/tutorials/simple_applications/pytorch/mnist/
pip install -r requirements.txt
cpython 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
```

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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
1. Import the TensorFlow IPU module

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

```python
from tensorflow.python import ipu
```
2. Preparing the dataset

- Make sure the sizes of the datasets are divisible by the batch size
  
  ```python
def make_divisible(number, divisor):
    return number - number % divisor
  ```

- Adjust dataset lengths

  ```python
  (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:

```python
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

```python
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

```python
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.
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 IPUs

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`.

```python
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
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.