

# DPU on PYNQ (3)

DNNDK

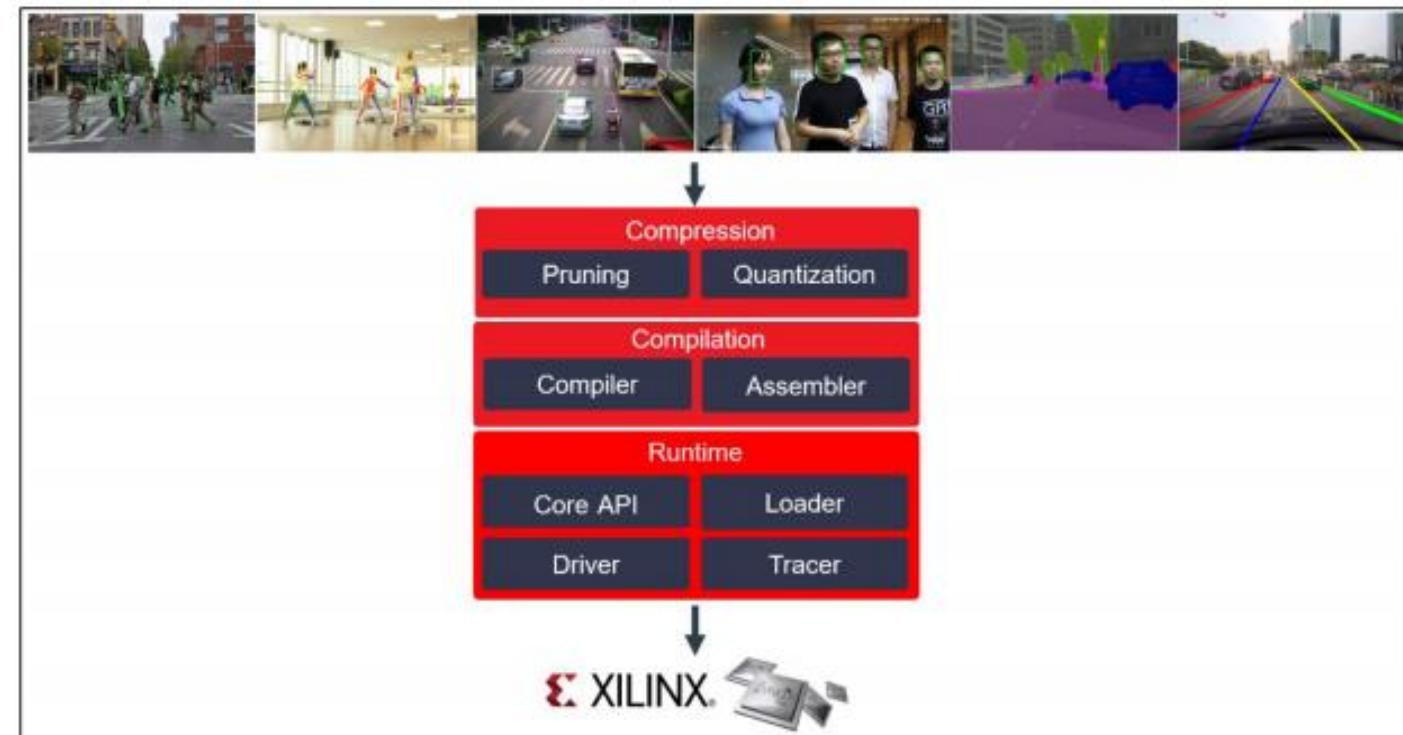
# Outline

- DNNDK
  - Overview
  - Toolchain Introduction
- DNNDK lab - Host
- DNNDK lab - Board

# DNNDK

Deep Neural Network Development Kit (DNNDK) is a full-stack deep learning SDK for the Deeplearning Processor Unit (DPU). It provides a unified solution for deep neural network inference applications by providing pruning, quantization, compilation, optimization, and run time support.

- A complete set of optimized tool chains, including compression, compilation and runtime.
- Lightweight C/C++ and Python programming APIs.
- Easy-to-use with gradual learning curve.

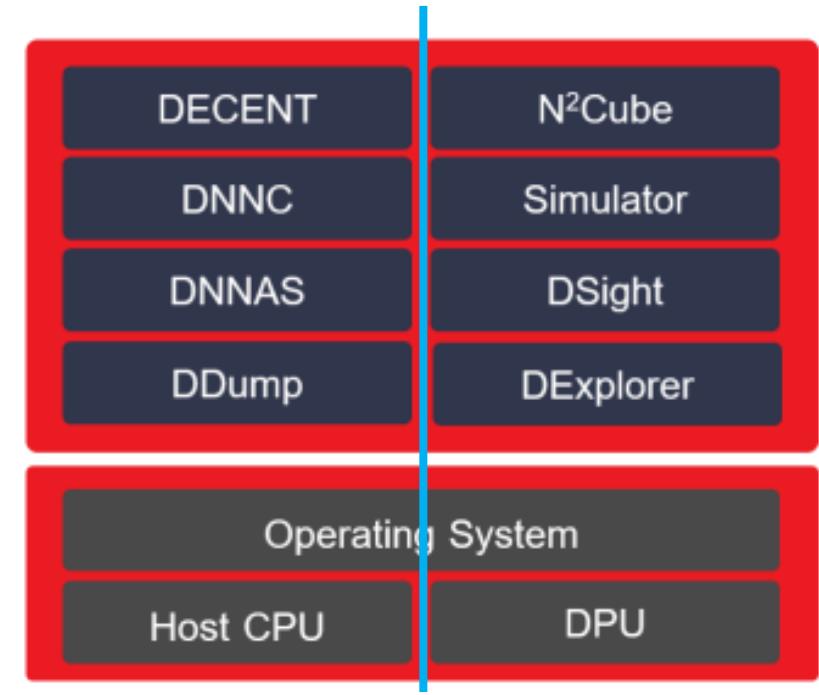


DNNDK framework

# DNNDK Framework

As shown in the figure below, DNNDK framework is composed of following components

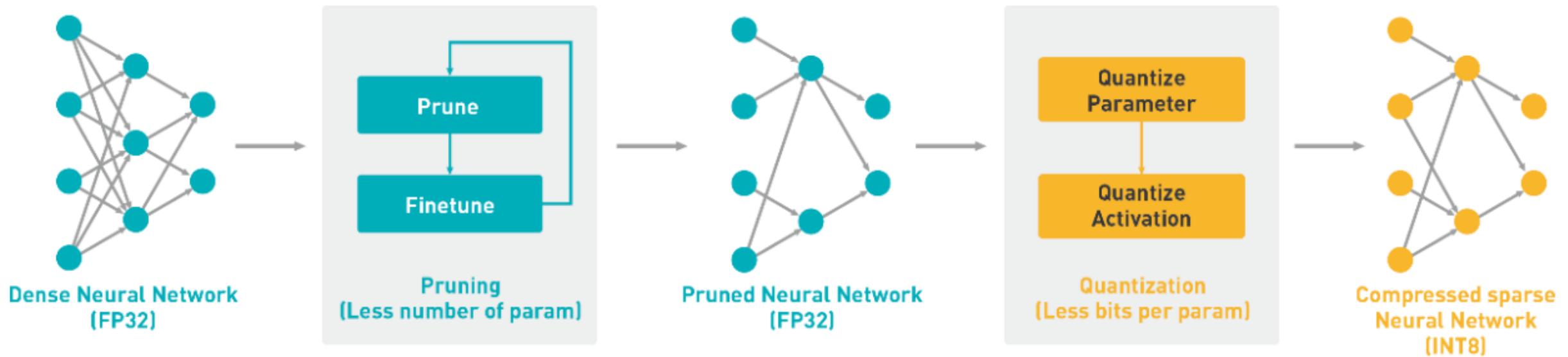
- Deep Compression Tool (DECENT)
- Deep Neural Network Compiler (DNNC)
- Deep Neural Network Assembler (DNNAS)
- Neural Network Runtime (N2Cube)
- DPU Simulator
- Profiler
- DExplorer
- DDump



DNNDK Toolchain

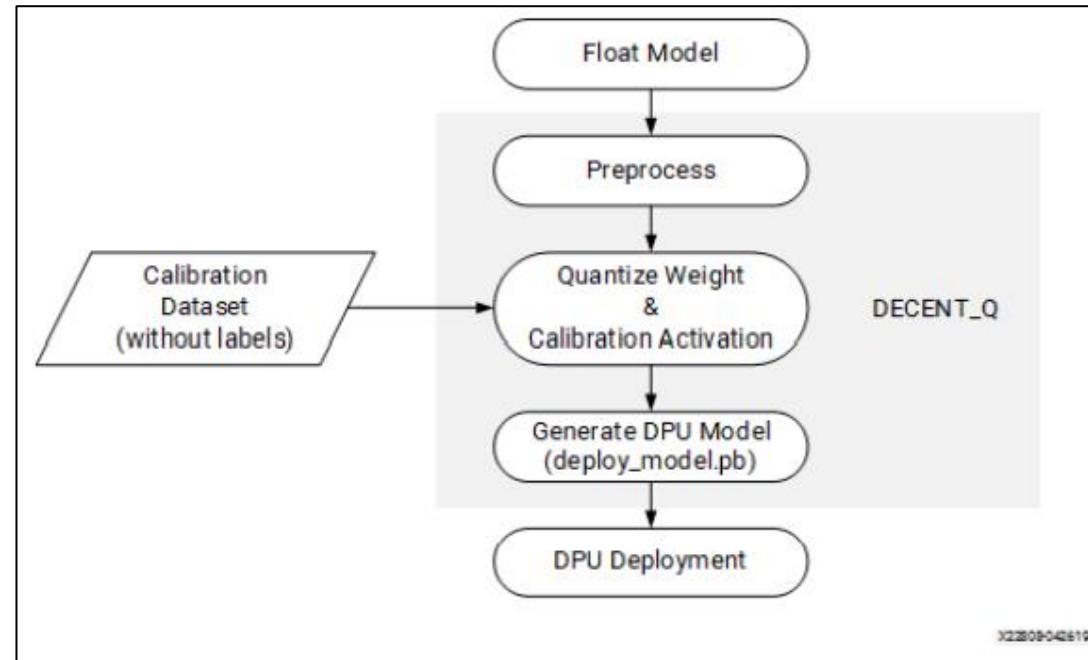
# DECENT

DECENT (Deep Compression Tool), employs coarse-grained pruning, trained quantization and weight sharing to address these issues while achieving high performance and high energy efficiency with very small accuracy degradation.



DECENT Pruning and Quantization Flow

# DECENT Workflow



DECENT Workflow

X2209042619

# DECENT Syntax

```
$decent_q quantize \
--input_frozen_graph ${frozen_graph_pb_file} \
--input_nodes ${input_nodes} \
--input_shapes ${input_shapes} \
--output_nodes ${output_nodes} \
--input_fn ${input_fn} \
[options]
```

```
decent_q quantize \
--input_frozen_graph frozen_resnet_v1_50.pb \
--input_nodes input \
--input_shapes ?,224,224,3 \
--output_nodes resnet_v1_50/predictions/Reshape_1 \
--input_fn resnet_v1_50_input_fn.calib_input \
--method 1 \
--gpu 0 \
--calib_iter 10 \
--output_dir ./quantize_results \
```

decent\_q syntax (tensorflow)

```
calib_image_dir = "../../calibration_data/images/"
calib_image_list = "../../calibration_data/calib.txt"
calib_batch_size = 50
def calib_input(iter):
    images = []
    line = open(calib_image_list).readlines()
    for index in range(0, calib_batch_size):
        curline = line[index * calib_batch_size + index]
        calib_image_name = curline.strip()
        image = cv2.imread(calib_image_dir + calib_image_name)
        image = central_crop(image, 224, 224)
        image = mean_image_subtraction(image, MEANS)
        images.append(image)
    return {"input": images}
```

decent\_q input\_fn

# DLet

- DLet is DNNDK host tool designed to parse and extract various DPU configuration parameters from DPU Hardware Handoff file HWH generated by Vivado.
- The usage info of DLet is shown below.

```
Usage: dlet <option>
Options are:
  -v  --version      Display version of DLet
  -f  --file        Specity hardware hand-off(HWH) file
  -h  --help         Display the usage of DLet
```

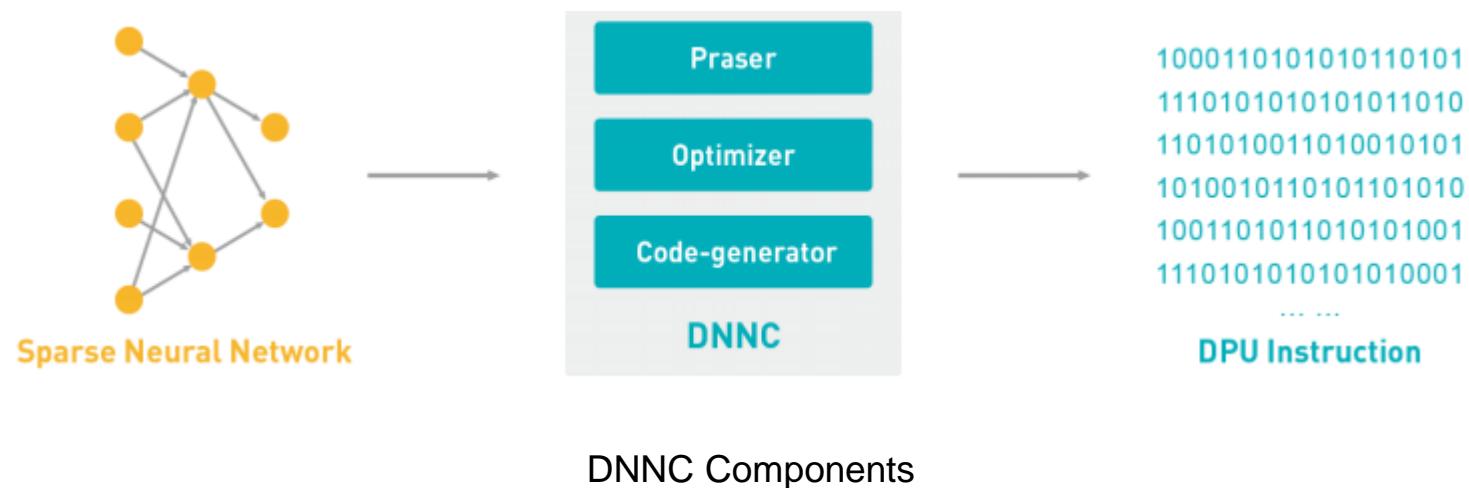
Dlet usage

Ex: **dlet -f ./pynqz2\_dpu.hwh**

Output: \*.dcf file

# DNNC

The architecture of the Deep Neural Network Compiler (DNNC) compiler is shown in the following figure. The front-end parser is responsible for parsing the Caffe/TensorFlow model and generates an intermediate representation (IR) of the input model. The optimizer handles optimizations based on the IR, and the code generator maps the optimized IR to DPU instructions.



# DNNC sample

```
#!/usr/bin/env bash

net="resnet50"
CPU_ARCH="arm64"
DNNC_MODE="debug"
dnndk_board="ZCU102"
dnndk_dcf="../..../dcl/ZCU102.dcf"

model_dir="decent_output"
output_dir="dnnnc_output"

if [ ! -d "$model_dir" ]; then
    echo "Can not found directory of $model_dir"
    exit 1
fi

[ -d "$output_dir" ] || mkdir "$output_dir"

echo "Compiling Network ${net}"
dnnc --prototxt=${model_dir}/deploy.prototxt \
    --caffemodel=${model_dir}/deploy.caffemodel \
    --output_dir=${output_dir} \
    --net_name=${net} \
    --dcf=${dnndk_dcf} \
    --mode=${DNNC_MODE} \
    --cpu_arch=${CPU_ARCH}
```

DNNC Compilation Script for Caffe ResNet-50

```
Compiling network: resnet50
[DNNC][Warning] layer [prob] (type: Softmax) is not supported in DPU, deploy it in CPU instead.

DNNC Kernel topology "resnet50_kernel_graph.jpg" for network "resnet50"
DNNC kernel list info for network "resnet50"
    Kernel ID : Name
                  0 : resnet50_0
                  1 : resnet50_1
    Kernel Name : resnet50_0
    -----
    Kernel Type : DPUKernel
                  Code Size : 1.28MB
                  Param Size : 24.35MB
                  Workload MACs : 3262.50MOPS
                  I0 Memory Space : 2.25MB
                  Mean Value : 104, 107, 123,
                  Node Count : 55
                  Tensor Count : 56
                  Input Node(s) (H*W*C)
                                  conv1(0) : 224*224*3
                  Output Node(s) (H*W*C)
                                  fc1000(0) : 1*1*1000
    Kernel Name : resnet50_1
    -----
    Kernel Type : CPUKernel
                  Input Node(s) (H*W*C)
                                  prob : 1*1*1000
                  Output Node(s) (H*W*C)
                                  prob : 1*1*1000
```

DNNC Compilation Log for Caffe ResNet-50

# DDump

- DDump is a utility tool introduced to dump the info encapsulated inside DPU ELF file or hybrid executable or DPU shared library. It can facilitate the users to analyze and debug various issues.
- DDump is available for both x86 Linux host and DNNDK evaluation boards. Its usage info is shown in the figure below.

```
DDump - Xilinx DNNDK utility to parse and dump DPU ELF file or
       DPU hybrid executable file
Usage: ddump <option>
At least one of the following switches must be given:
-f  --file      Specify DPU hybrid executable or DPU ELF object
-k  --klist     Display each kernel general info from DPU ELF file
               or DPU hybrid executable file
-d  --dpu       Display DPU architecture info for each kernel
-c  --compiler  Display the DNCC compiler version for each kernel
-a  --all        Display all above info
-v  --version   Display DDump version info
-h  --help      Display this help info
```

DDump Usage Options

# DDump

```
ddump -f dpu_resnet50_0.elf -k
```

```
DPU Kernel List from file dnnc_output/dpu_resnet50_0.elf
    ID: Name
    0: resnet50_0
```

```
DPU Kernel name: resnet50_0
```

```
-> DPU Kernel general info
    Mode: NORMAL
    Code Size: 1.28MB
    Param Size: 24.35MB
    Workload MACs: 7358.50MOPS
    IO Memory Space: 2.25MB
        Mean Value: 104, 107, 123
        Node Count: 55
        Tensor Count: 56
    Tensor In(H*W*C)
        Tensor ID-0: 224*224*3
    Tensor Out(H*W*C)
        Tensor ID-55: 1*1*1000
```

DDump DPU Kernel Info for ResNet50

```
ddump -f dpu_resnet50_0.elf -d
```

```
DPU Kernel List from file dpu_resnet50_0.elf
    ID: Name
    0: resnet50_0
```

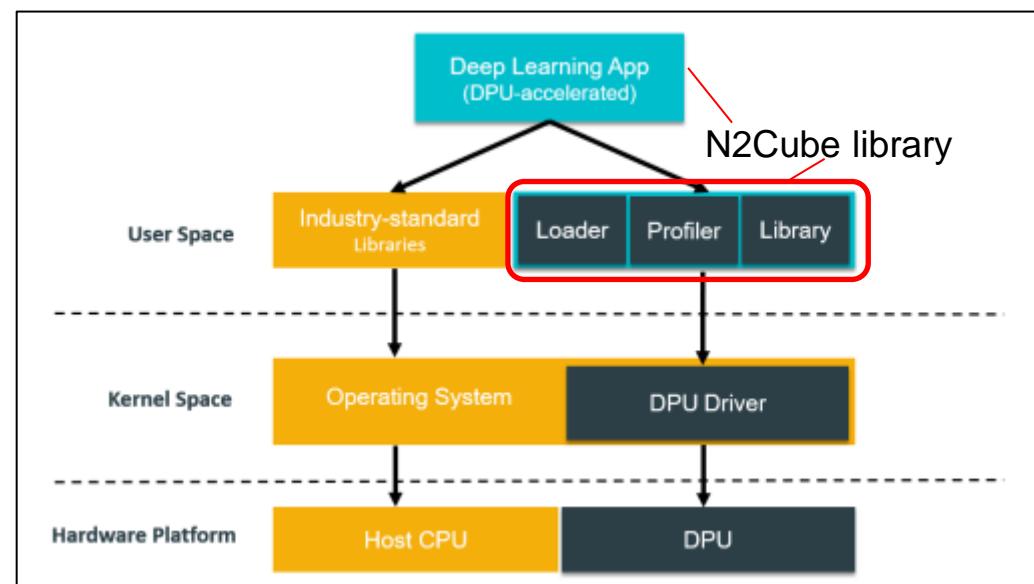
```
DPU Kernel name: resnet50_0
```

```
-> DPU architecture info
    DPU ABI Ver: v2.0
DPU Configuration Parameters
    DPU Target Ver: 1.4.0
    DPU Arch Type: B512
    RAM Usage: high
    DepthwiseConv: Enabled
    DepthwiseConv+Relu6: Enabled
    Conv+Leakyrelu: Enabled
    Conv+Relu6: Enabled
    Channel Augmentation: Enabled
    Average Pool: Disabled
```

DDump DPU Arch Info for ResNet50

# N2Cube

- The Cube of Neutral Networks (N2Cube) is the DPU runtime engine.
- It acts as the loader for the DNNDK applications and handles resource allocation and DPU scheduling.
- Its core components include DPU driver, DPU loader, tracer, and programming APIs for application development.



# DExplorer

- DExplorer is a utility running on the target board.
- It provides DPU running mode configuration, DNNDK version checking, DPU status checking, and DPU core signature checking.
- The following figure shows the help information about the usage of DExplorer.

```
Usage: dexplorer <option>
Options are:
  -v  --version    Display version info for each DNNDK component
  -s  --status     Display the status of DPU cores
  -w  --whoami     Display the info of DPU cores
  -m  --mode       Specify DNNDK N2Cube running mode: normal, profile, or debug
  -t  --timeout   Specify DPU timeout limitation in seconds under integer range of [1, 100]
  -h  --help        Display this information
```

DExplorer Usage Options

# DExplorer

```
root@dp-n1:~# dexplorer -s
[DPU cache]
Enabled

[DPU mode]
normal

[DPU timeout limitation (in seconds)]
5

[DPU Debug Info]
Debug level      : 9
Core 0 schedule : 0
Core 0 interrupt: 0

[DPU Resource]
DPU Core        : 0
State           : Idle
PID              : 0
TaskID          : 0
Start            : 0
End              : 0

[DPU Registers]
VER              : 0x05c1c6bd
RST              : 0x000000ff
ISR              : 0x00000000
IMR              : 0x00000000
IRSR             : 0x00000000
ICR              : 0x00000000

DPU Core        : 0
HP_CTL          : 0x07070f0f
ADDR_IO          : 0x00000000
ADDR_WEIGHT      : 0x00000000
ADDR_CODE        : 0x00000000
ADDR_PROF        : 0x00000000
```

DExplorer Status

```
root@xilinx-zcu102-2019_1:~$dexplorer -w
[DPU IP Spec]
IP   Timestamp          : 2019-07-24 11:15:00
DPU Core Count         : 3

[DPU Core Configuration List]
DPU Core               : #0
DPU Enabled            : Yes
DPU Arch               : B4096
DPU Target Version    : v1.4.0
DPU Frequency          : 325 MHz
Ram Usage              : Low
DepthwiseConv          : Enabled
DepthwiseConv+Relu6    : Enabled
Conv+Leakyrelu         : Enabled
Conv+Relu6              : Enabled
Channel Augmentation  : Enabled
Average Pool           : Enabled

DPU Core               : #1
```

Sample DPU Signature with Configuration Parameters

# DSight

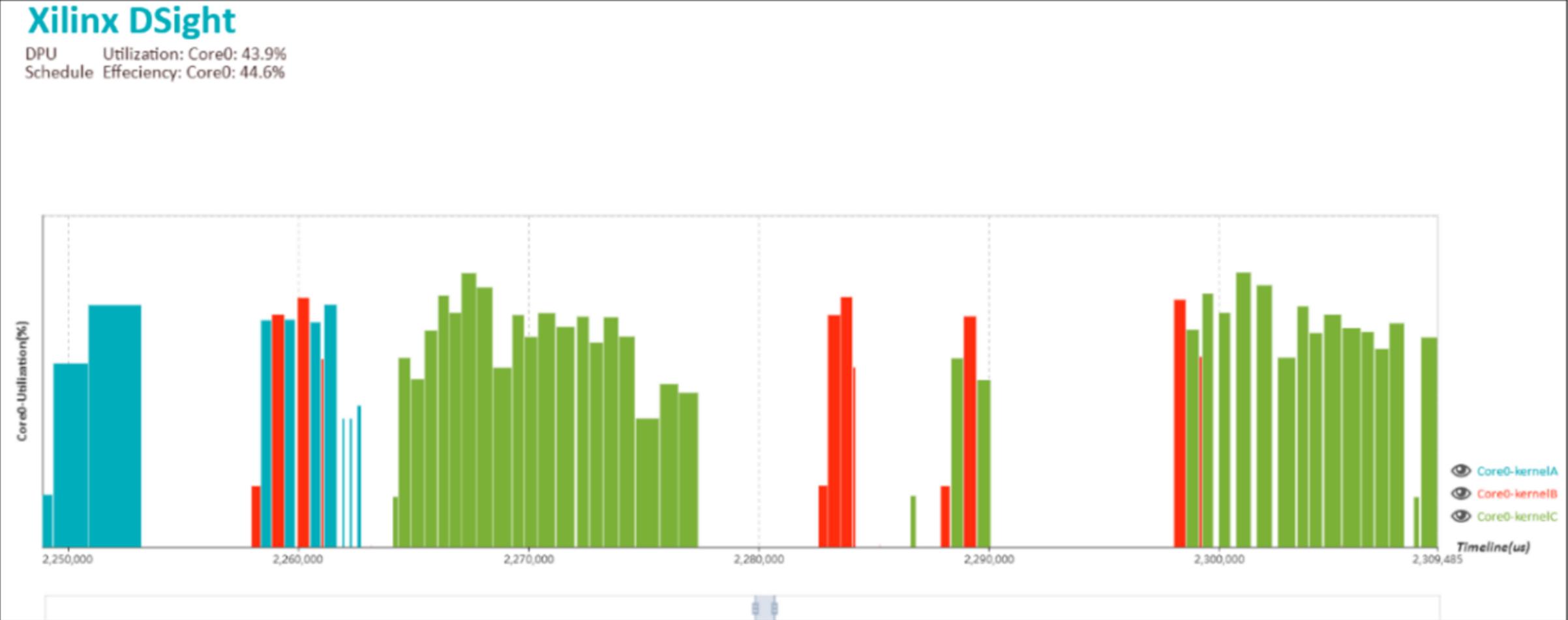
DSight is the DNNDK performance profiling tool. It is a visual performance analysis tool for neural network model profiling.

## Dsight profiling Step:

1. Set N2Cube to profile mode using the command *dexplorer -m profile*
2. Run the deep learning application. When finished, a profile file with the name *dpu\_trace\_[PID].prof* is generated. (PID is the process ID of the deep learning application).
3. Generate the html file with the DSight tool using the command: *dsight -p dpu\_trace\_[PID].prof*. An html file with the name *dpu\_trace\_[PID].html* is generated.
4. Open the generated html file with web browser.

```
root@xlnx:~# dsight -h
Usage: dsight <option>
Options are:
  -p  --profile    Specify DPU trace file for profiling
  -v  --version    Display DSight version info
  -h  --help        Display this information
```

# DSight Profiling Charts



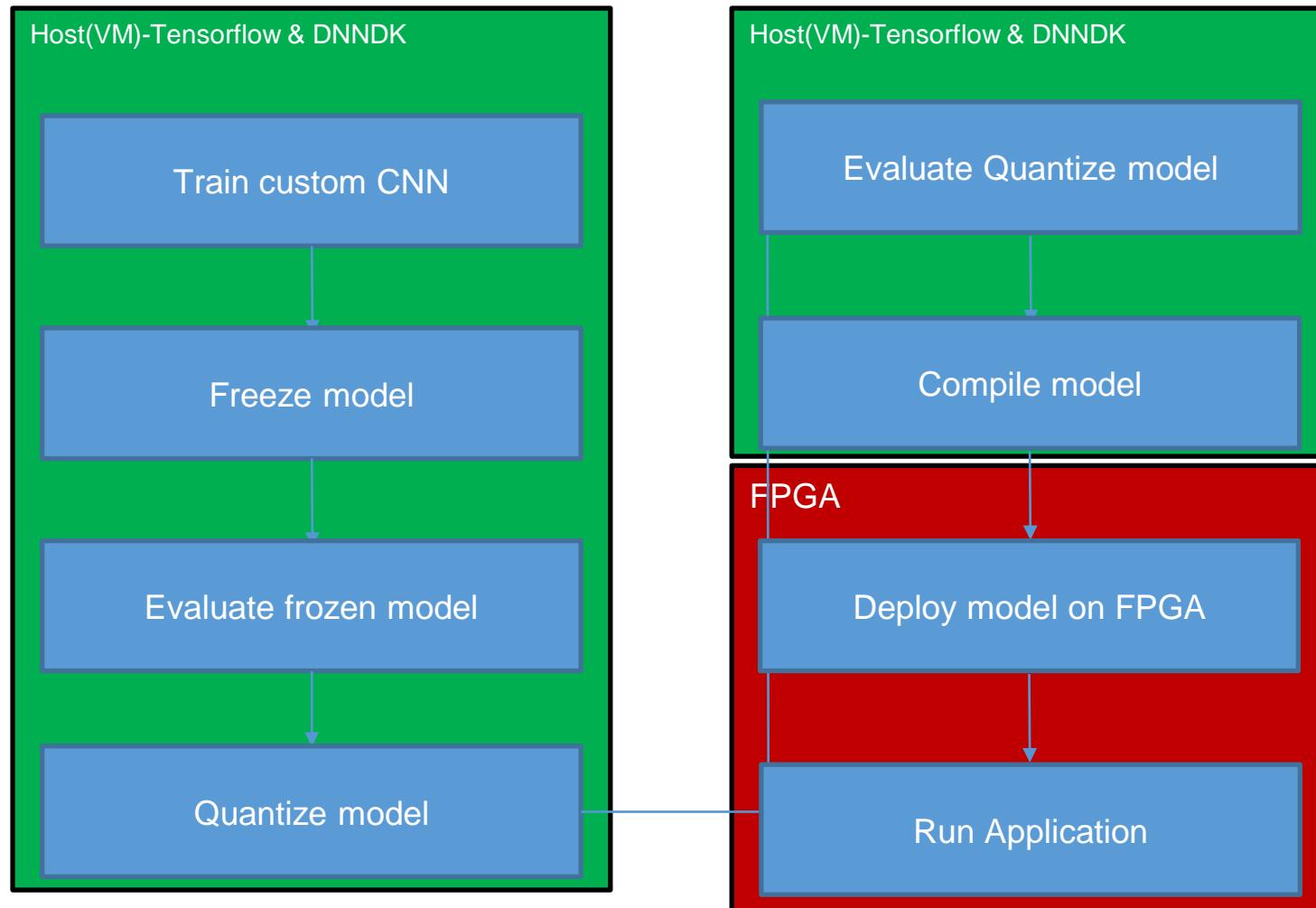
# Xilinx Vitis-ai AI-Model-Zoo

Vitis-ai AI-Model-Zoo <https://github.com/Xilinx/Vitis-AI/tree/master/models/AI-Model-Zoo>

Example code for each model: <https://github.com/Xilinx/Vitis-AI/tree/master/models/AI-Model-Zoo/caffe-xilinx/examples>



# Today's Lab



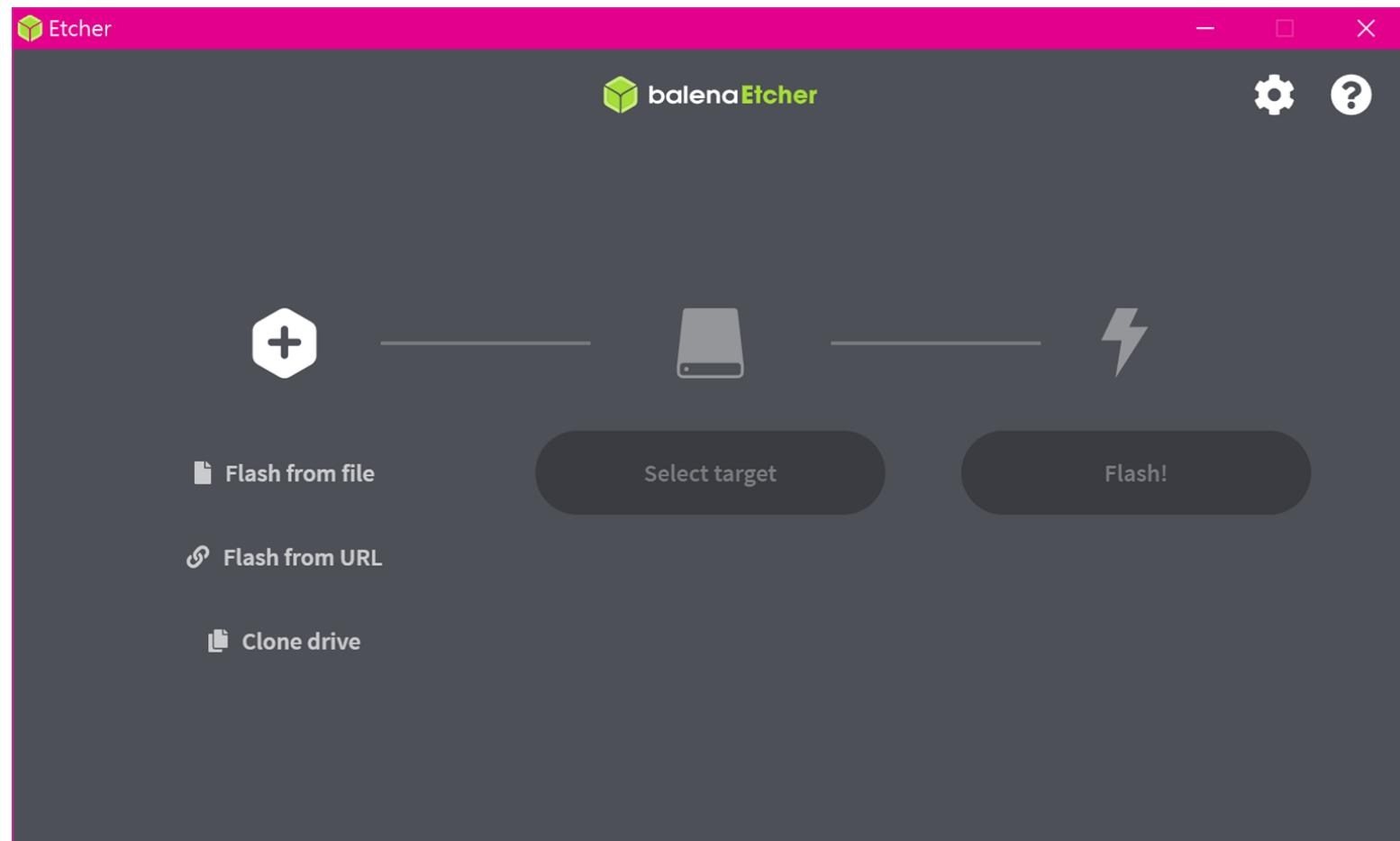
# Flash Boot Image

# Mount SD card and open Etcher

Path: Desktop(桌面)/Advanced\_FPGA\_Design/EtcherPortable/

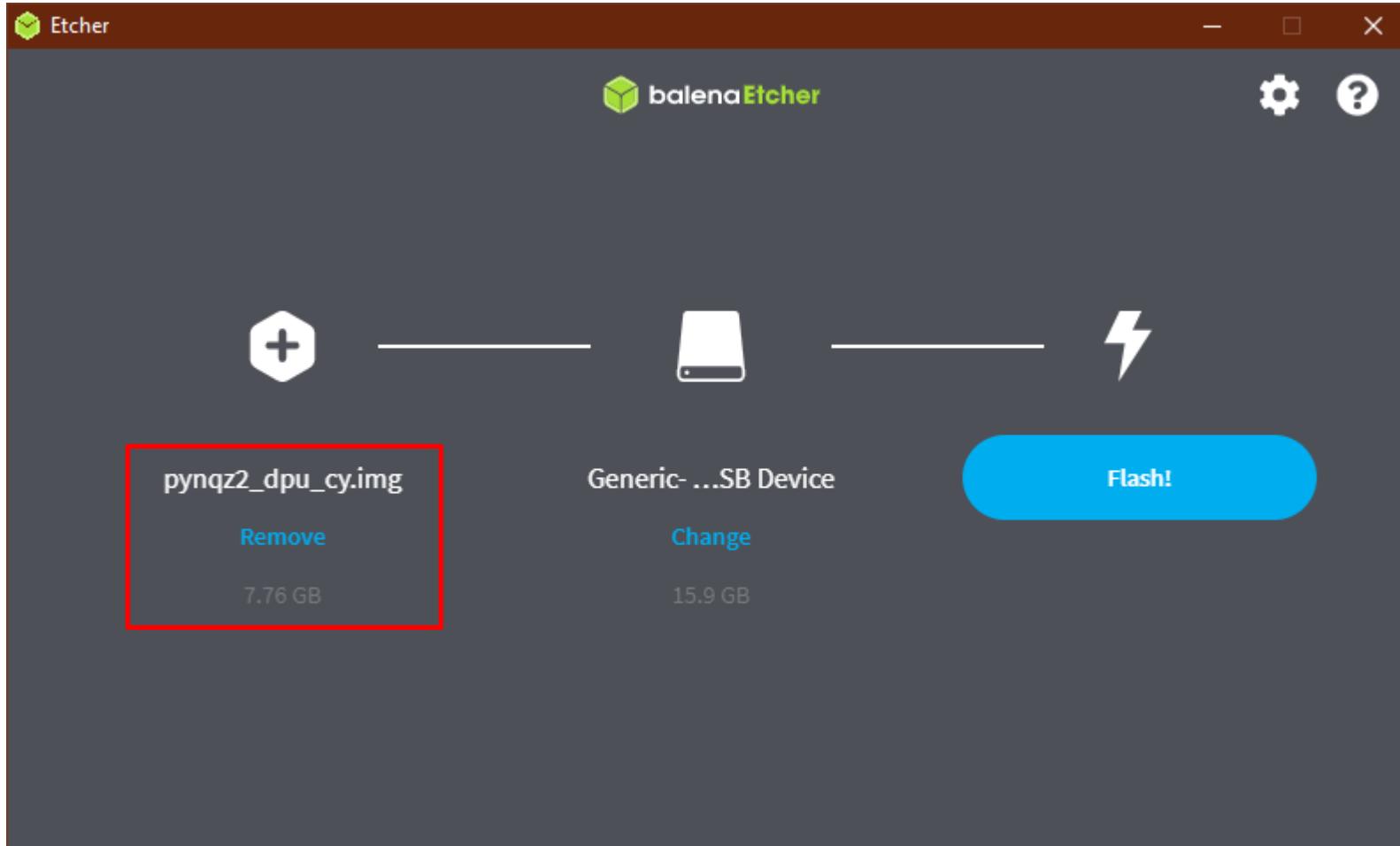


請記得把小張的SD卡插進轉接卡後  
再把整張轉卡插進電腦裡



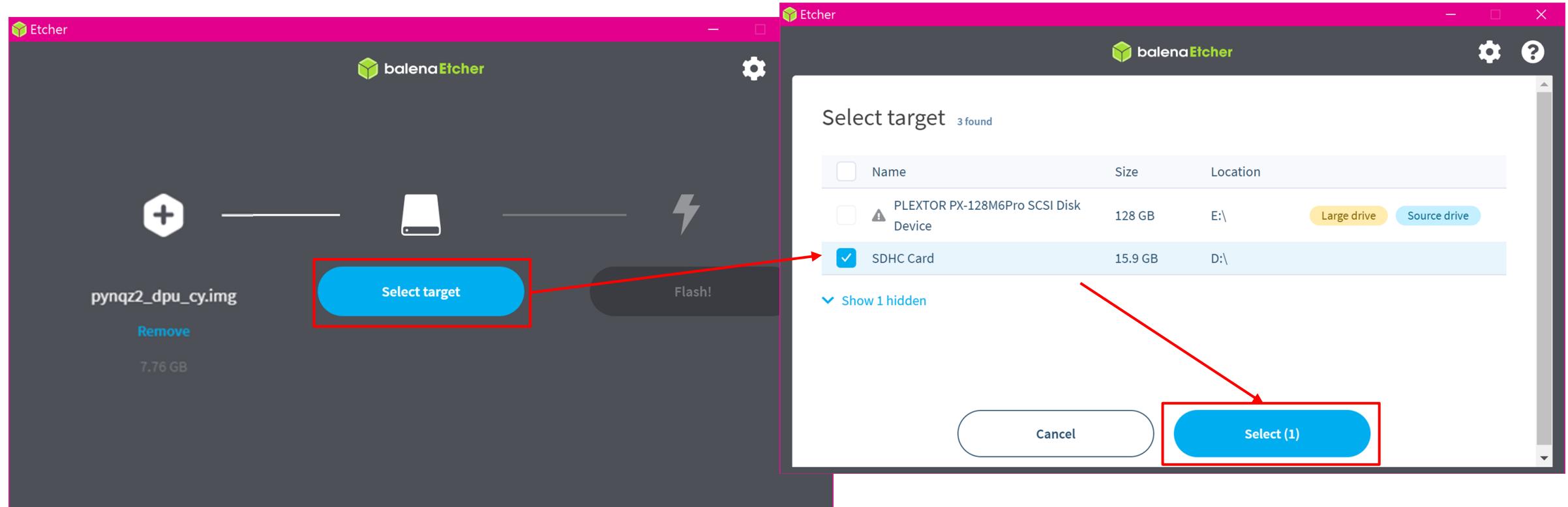
# Select img file

select pynqz2\_dpu\_cy\_v2.img , [Download link](#)



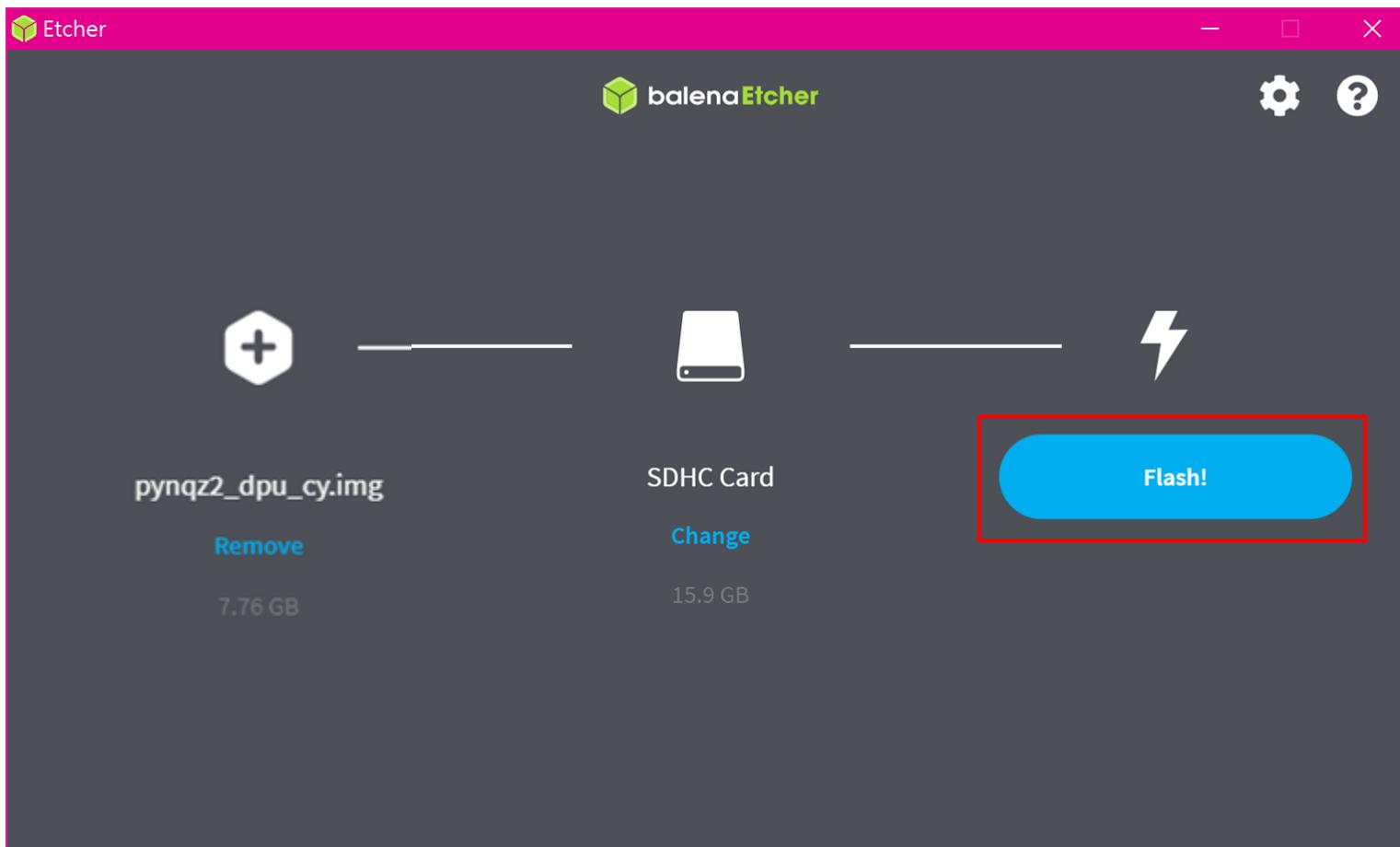
# Select Target

select SDHC

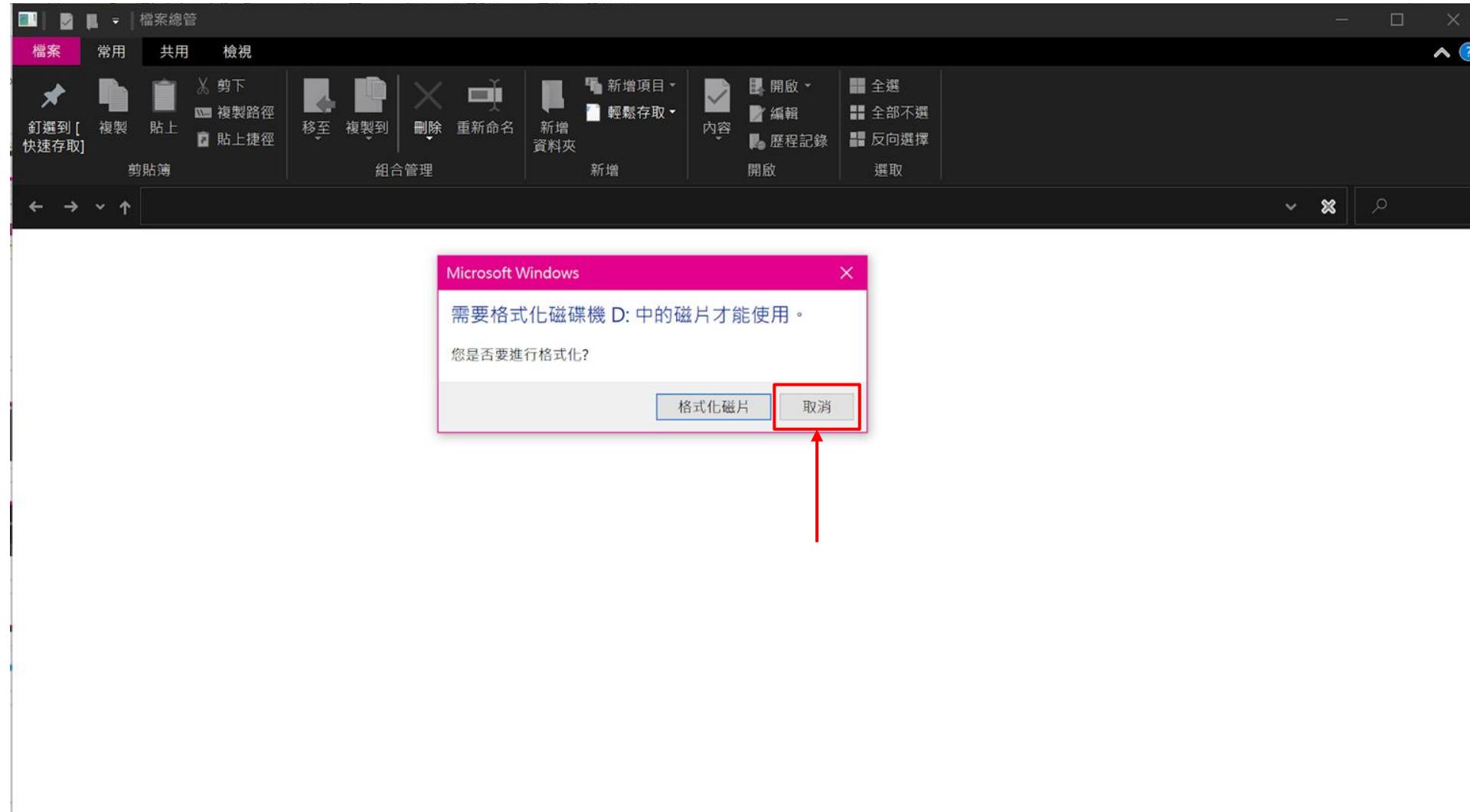


# Click Flash

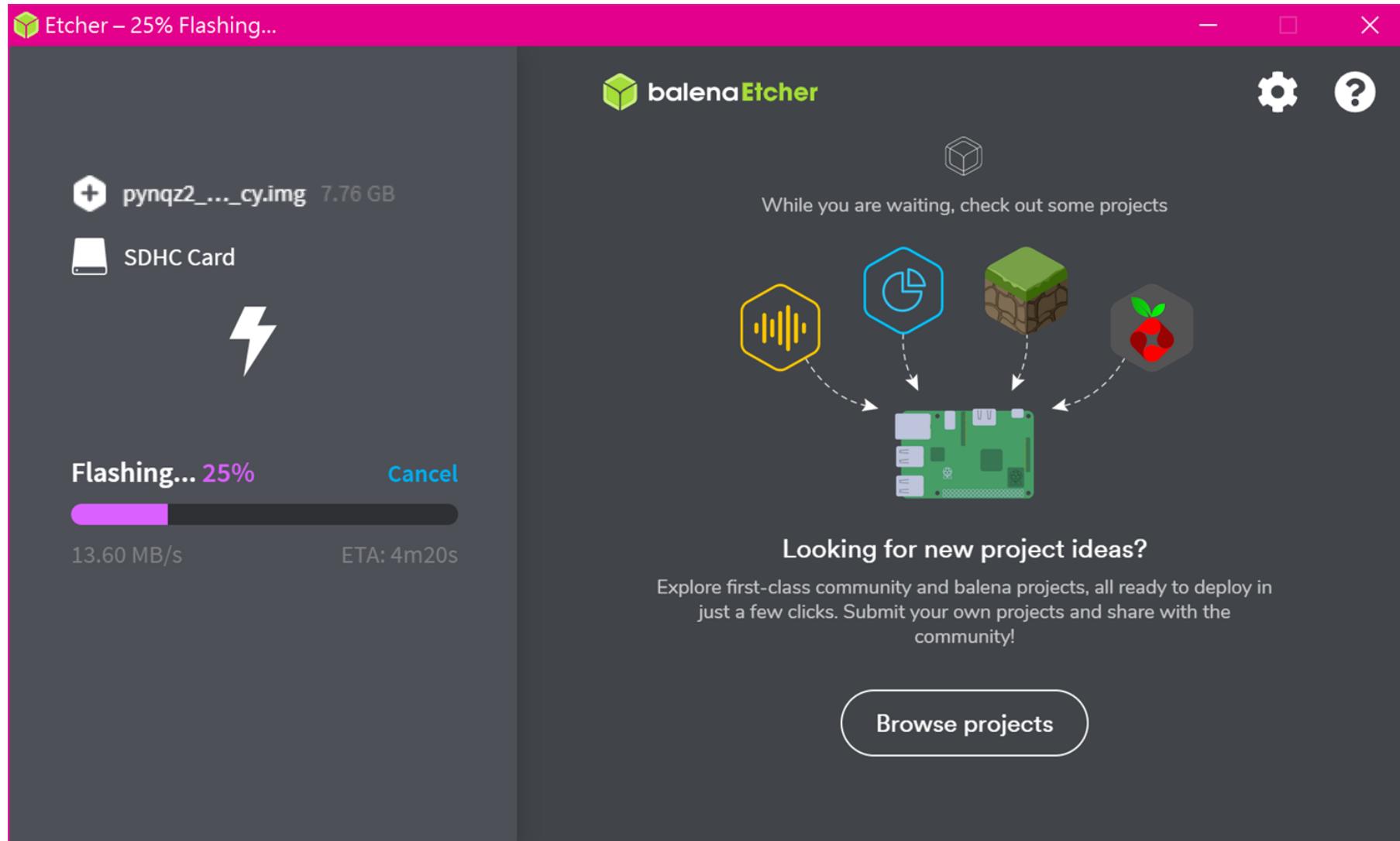
Please make sure your etcher looks like this



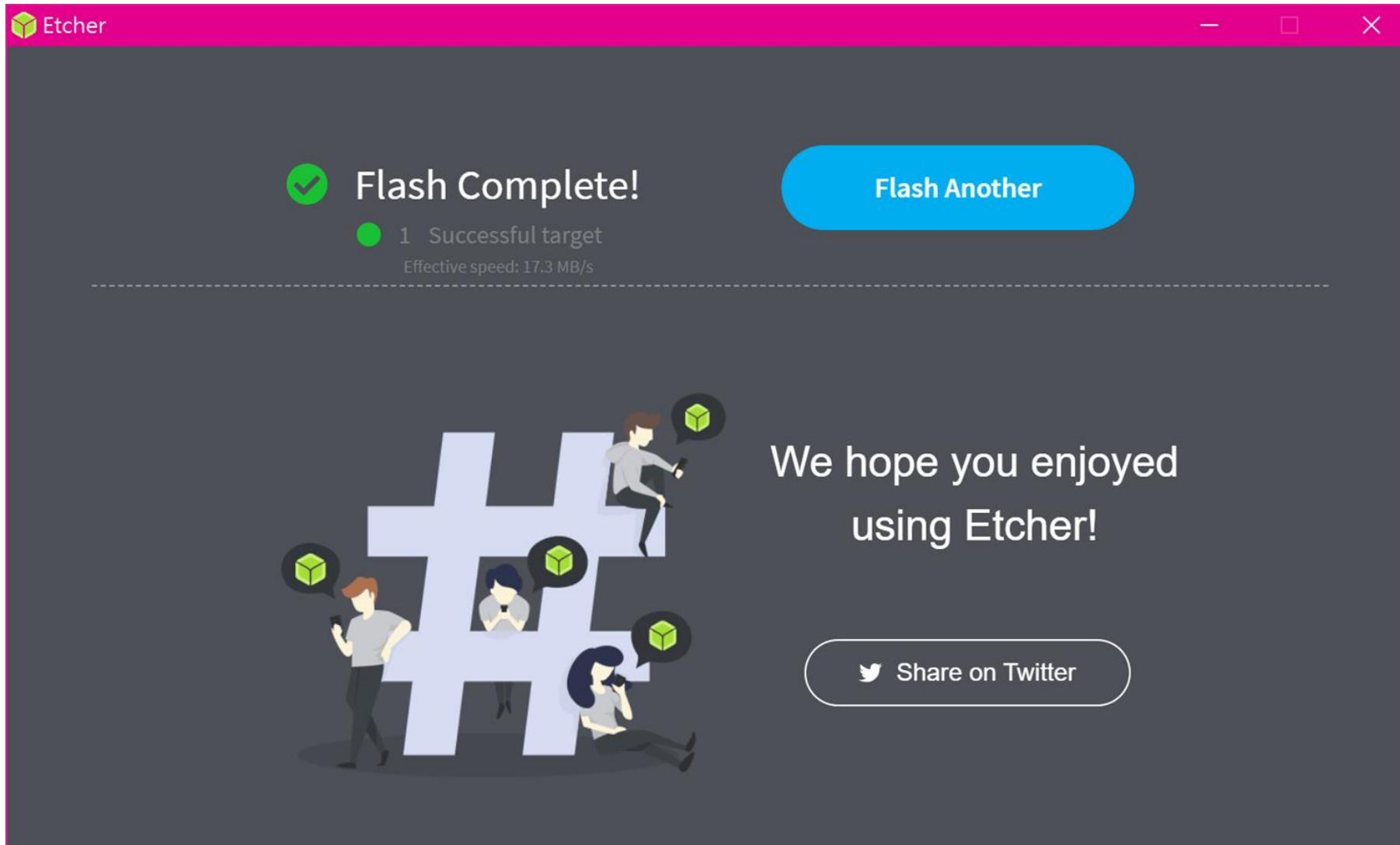
Select cancel and ignore error message if this pop out.  
**Do not format the disk manually!**



# Make sure it's flashing and wait for it



# Done!

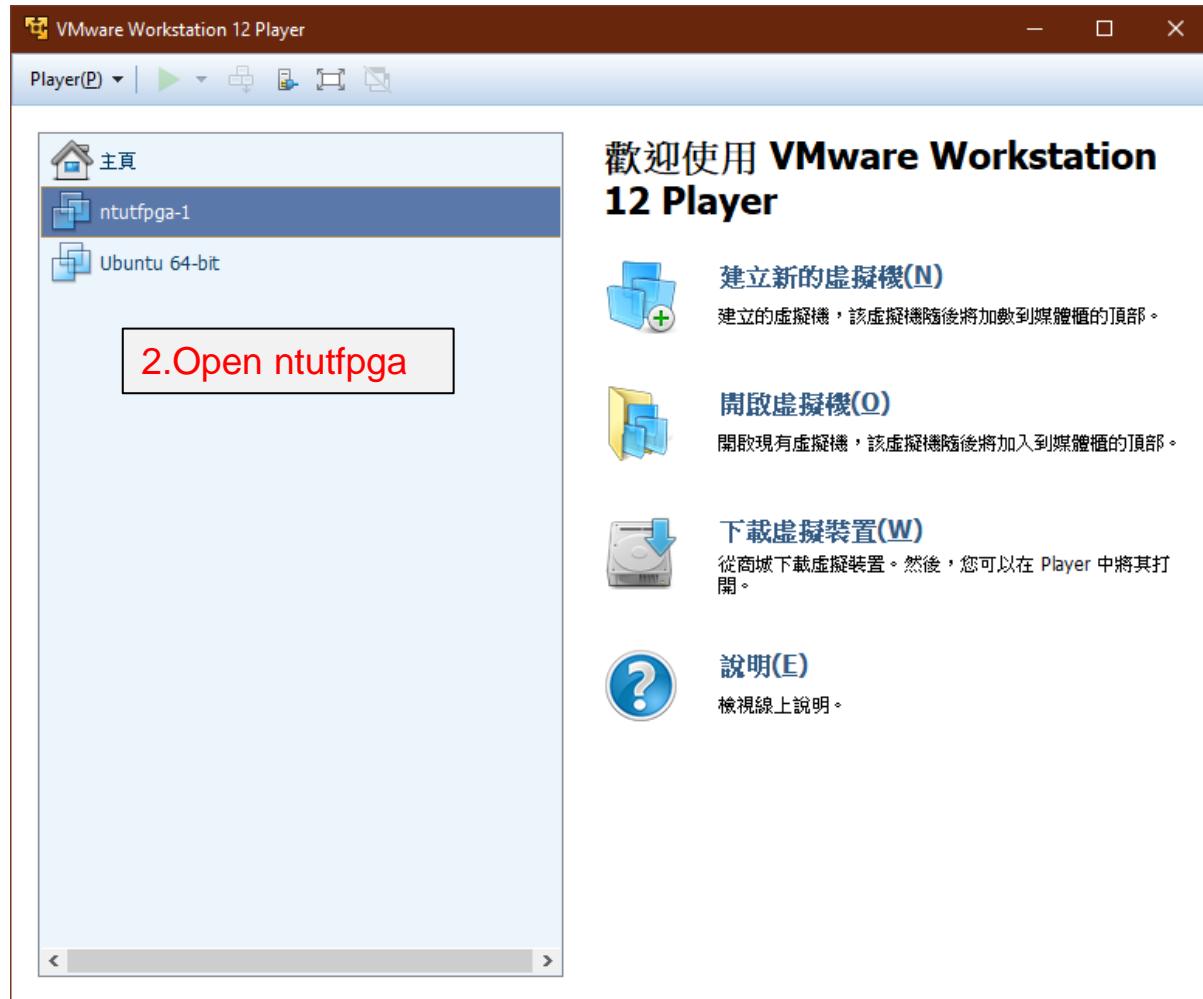


DNNDK Lab - host

# Open VMware



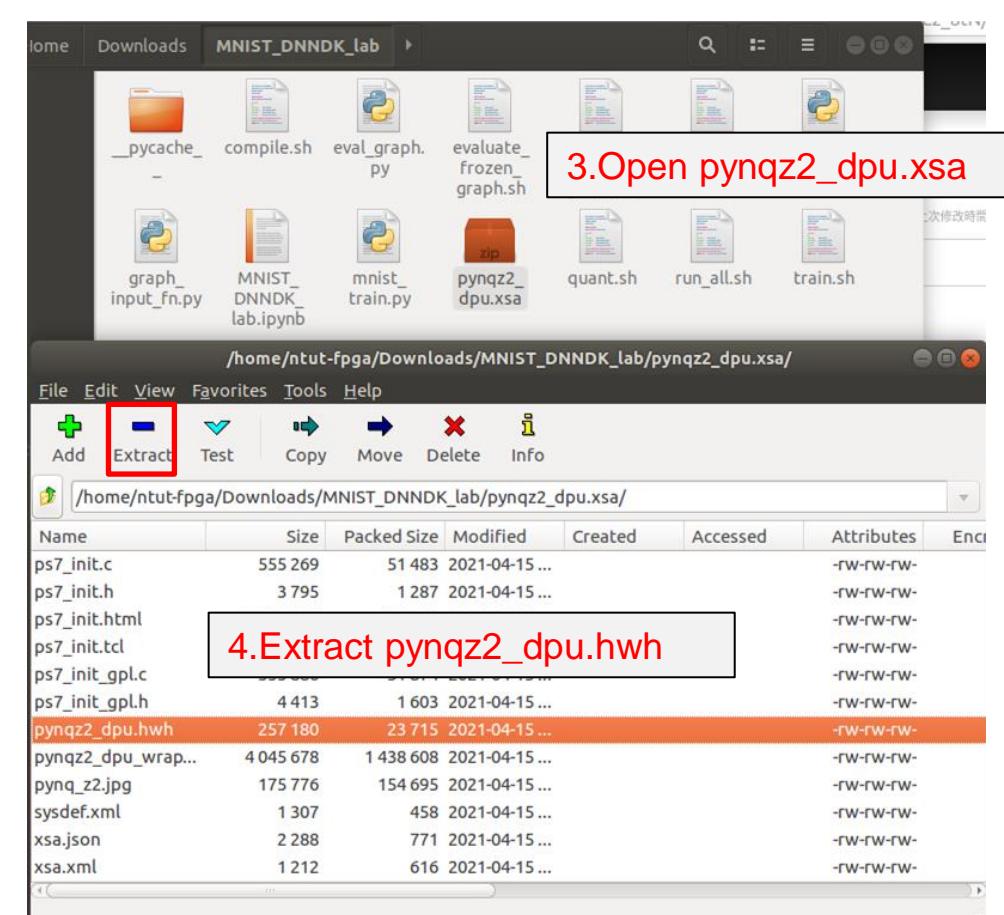
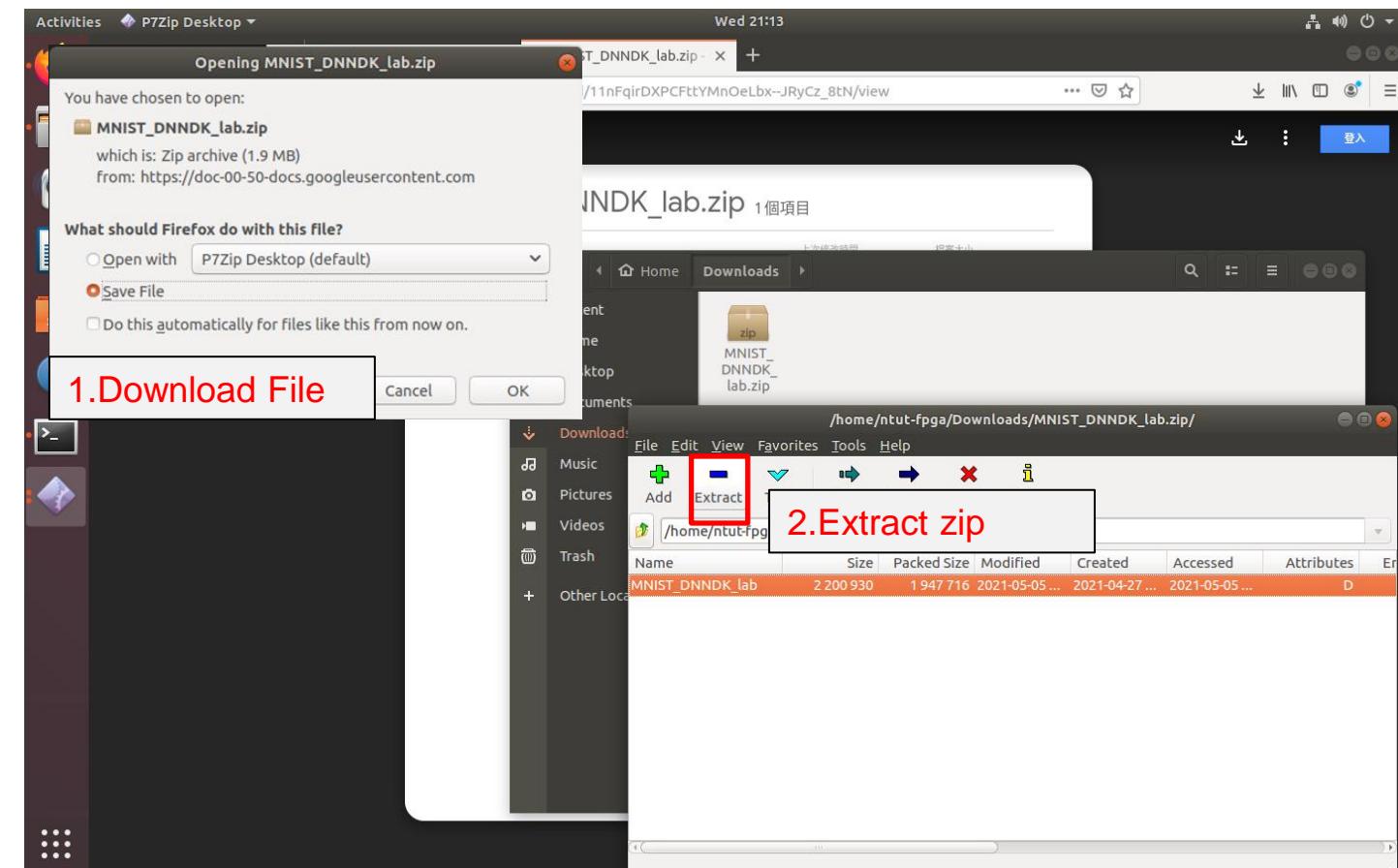
1.Open VMware



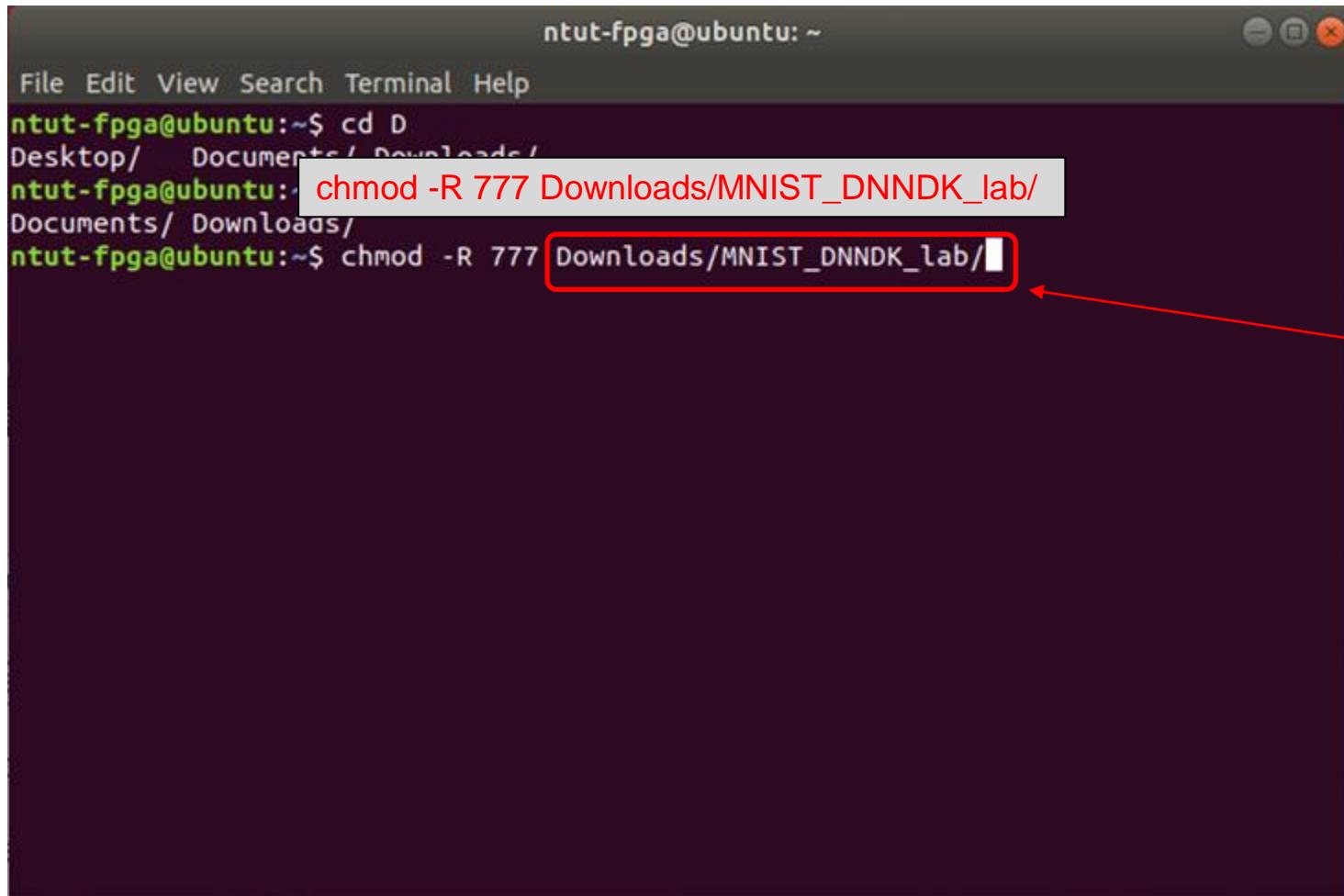
Password:ntutfpga

# Download lab file on VM

Lab file Download Link:[https://drive.google.com/file/d/11nFqirDXPCFttYMnOeLbx--JRyCz\\_8tN/view?usp=sharing](https://drive.google.com/file/d/11nFqirDXPCFttYMnOeLbx--JRyCz_8tN/view?usp=sharing)



# Change lab file permission



A screenshot of a terminal window titled "ntut-fpga@ubuntu: ~". The window shows the following command sequence:

```
ntut-fpga@ubuntu:~$ cd D  
Desktop/ Documents/ Downloads/  
ntut-fpga@ubuntu:~$ chmod -R 777 Downloads/MNIST_DNNDK_lab/  
Documents/ Downloads/  
ntut-fpga@ubuntu:~$ chmod -R 777 Downloads/MNIST_DNNDK_lab/
```

The last command, "chmod -R 777 Downloads/MNIST\_DNNDK\_lab/", is highlighted with a red rectangle. A red arrow points from the explanatory text below to this highlighted command.

Directory of the lab file that you just downloaded and extracted

# Run Jupyter notebook

```
python3 -c "from notebook.auth import passwd; print(passwd('', algorithm='sha1'))"  
-> sha1:a46679d19a61:58b4dd703239e6de7f445dff05a8407640d19dcb  
jupyter-lab --ServerApp.ip="*" --ServerApp.password="sha1:a46679d19a61:58b4dd703239e6de7f445dff05a8407640d19dcb"
```

```
ntut-fpga@ubuntu:~$ python3 -c "from notebook.auth import passwd; print(passwd('', algorithm='sha1'))"
```

```
sha1:92e29861c388:3bb89b37d2022aff96c3c4798f11e2171d26056a
```

```
ntut-fpga@ubuntu:~$ jupyter-lab --ServerApp.ip="*" --ServerApp.password="sha1:92e29861c388:3bb89b37d2022aff96c3c4798f11e2171d26056a"
```

(New terminal)

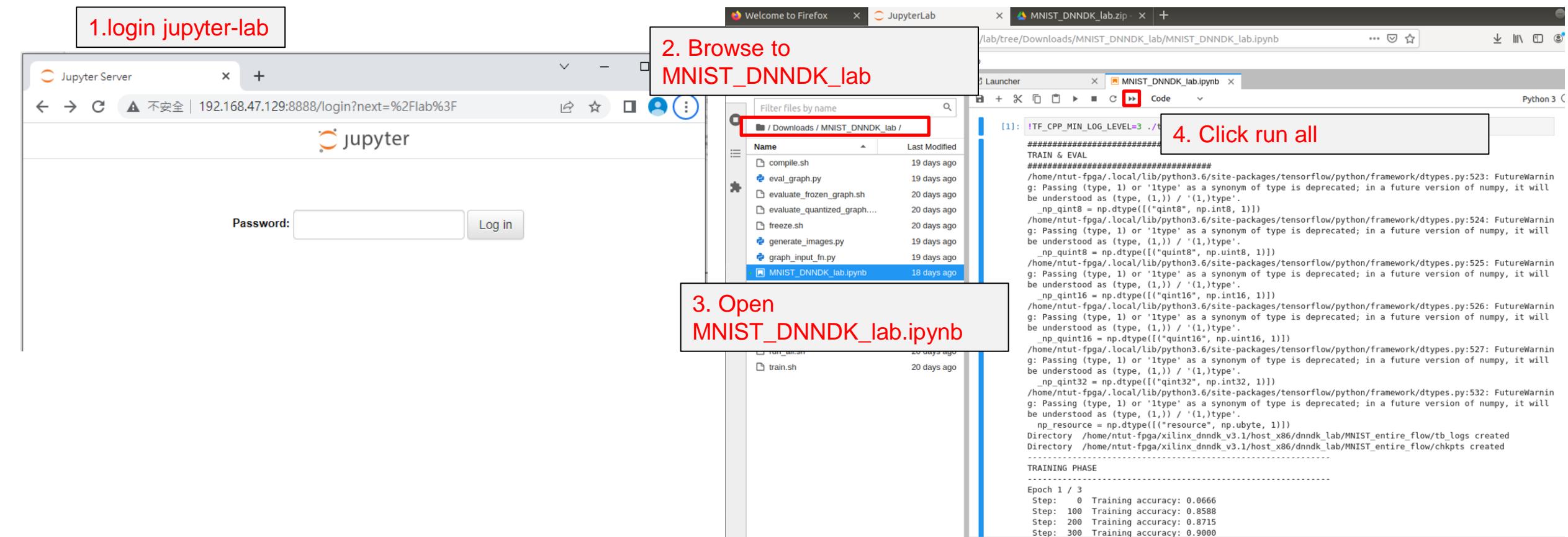
ifconfig

```
ntut-fpga@ubuntu:~$ ifconfig  
docker0: flags=4099<UP,BROADCAST,MULTICAST> mtu 1500  
      inet 172.17.0.1 netmask 255.255.0.0 broadcast 172.17.255.255  
        ether 02:42:32:1e:2e:0b txqueuelen 0 (Ethernet)  
          RX packets 0 bytes 0 (0.0 B)  
          RX errors 0 dropped 0 overruns 0 frame 0  
          TX packets 0 bytes 0 (0.0 B)  
          TX errors 0 dropped 0 overruns 0 carrier 0 collisions 0
```

Browse [192.168.47.129:8888](http://192.168.47.129:8888) in windows browser and log in.

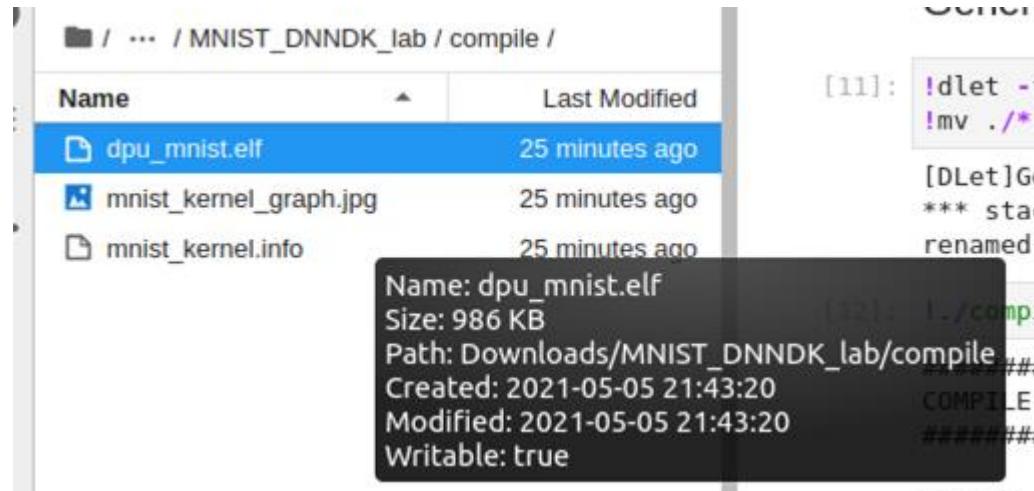
```
ens33: flags=4163<UP,BROADCAST,RUNNING,MULTICAST> mtu 1500  
      inet 192.168.47.129 netmask 255.255.255.0 broadcast 192.168.47.255  
        inet6 fe80::dce2:90f8:63c0:37ac prefixlen 64 scopeid 0x20<link>  
          ether 00:0c:29:14:ba:ca txqueuelen 1000 (Ethernet)  
            RX packets 640006 bytes 942999267 (942.9 MB)  
            RX errors 0 dropped 0 overruns 0 frame 0  
            TX packets 102652 bytes 7605296 (7.6 MB)  
            TX errors 0 dropped 0 overruns 0 carrier 0 collisions 0
```

# run Jupyter-lab

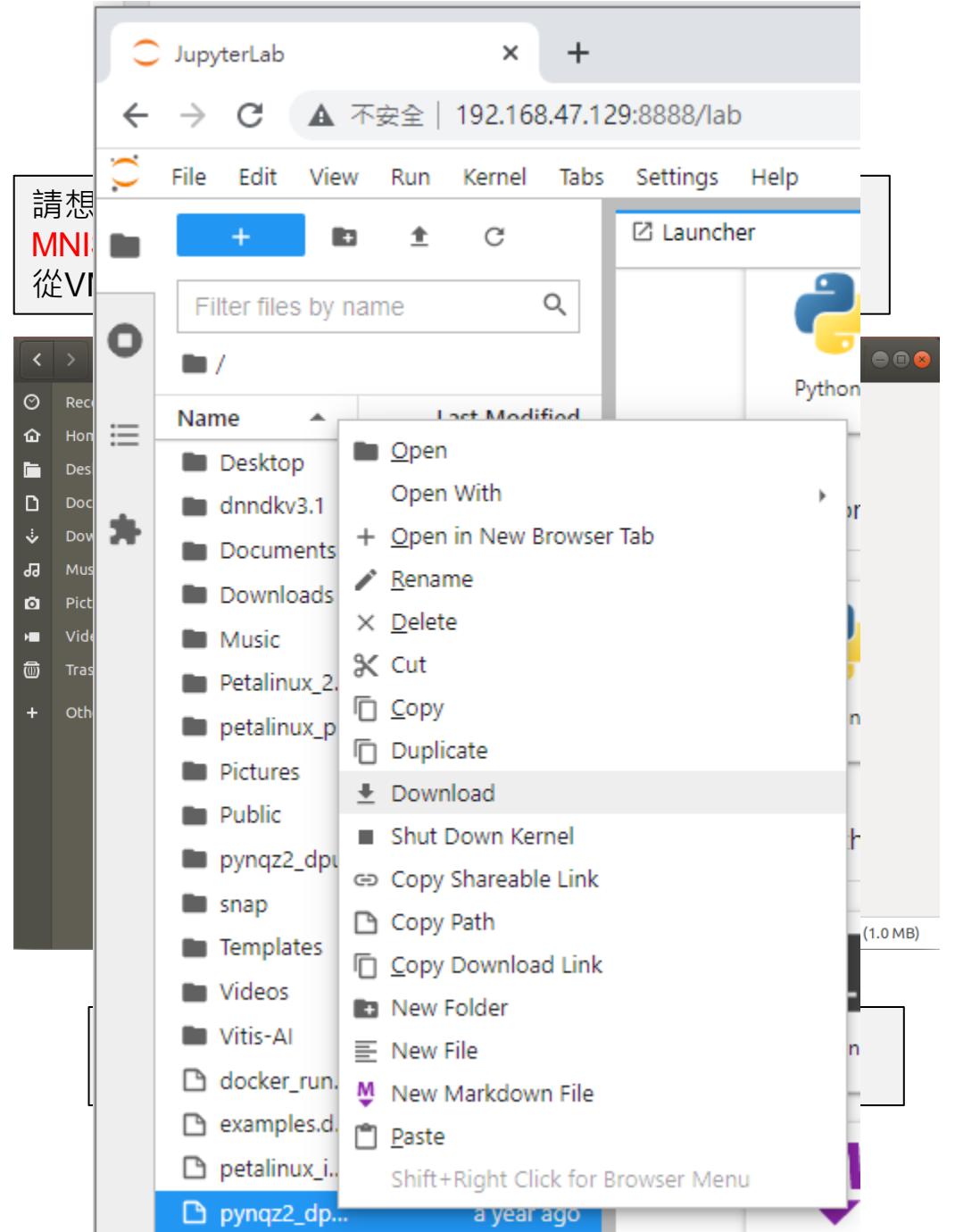


# Copy mnist.elf from VM

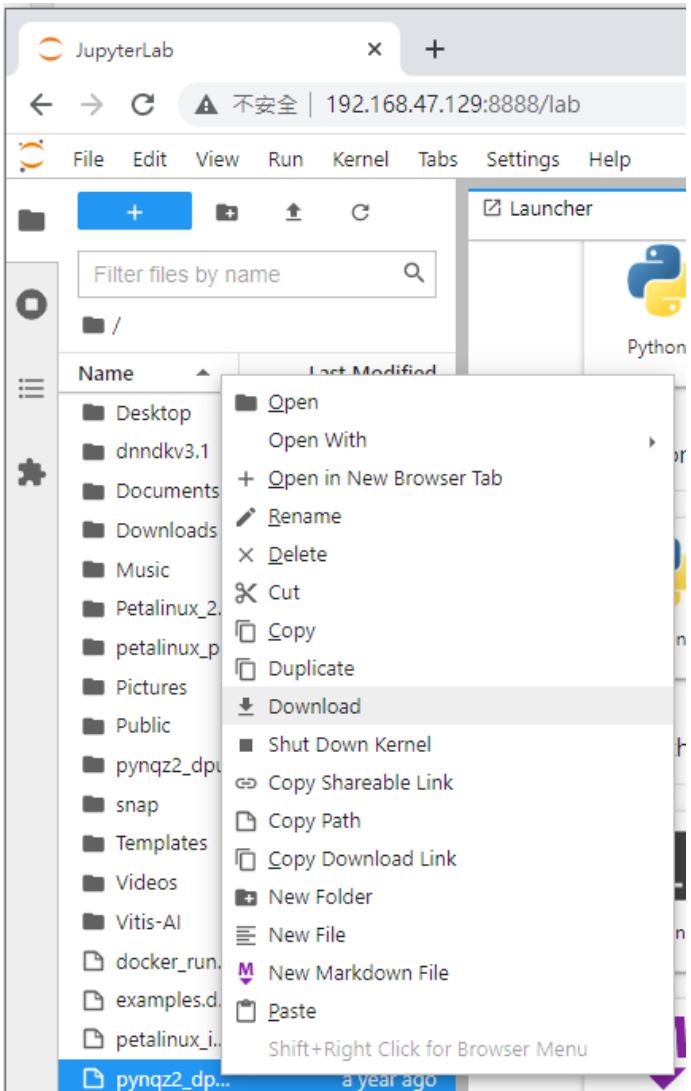
Trying to copy  
MNIST\_DNNDK\_lab/compile/dpu\_mnist.elf  
from VM to Windows !



Do not use ctrl + c or usb  
dirve. Please use cloud drive  
or other ways through internet.



# Copy mnist.elf from VM



Example

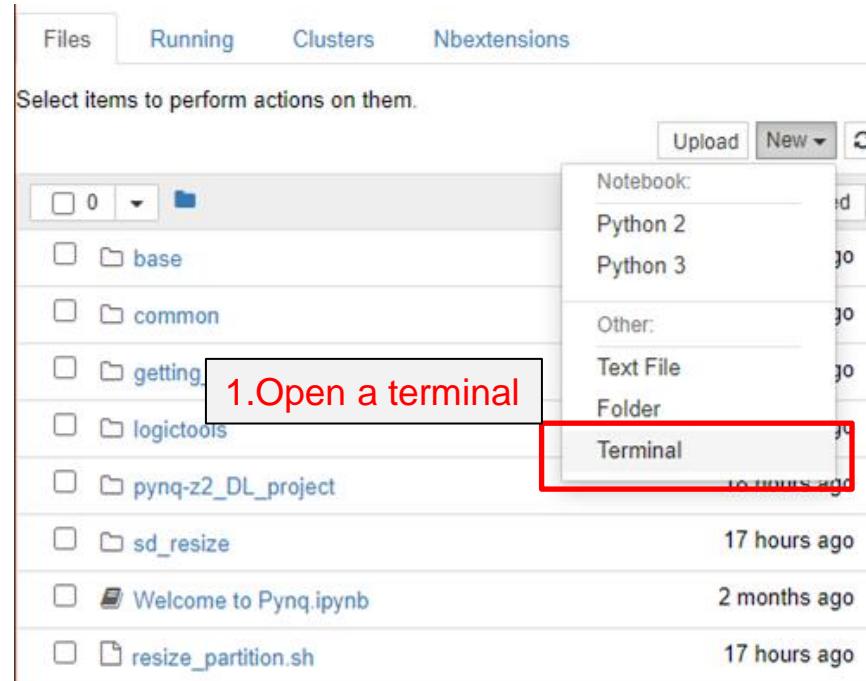
若是使用windows訪問VM中的jupyter-lab，可直接用download功能將檔案透過瀏覽器下載到windows環境中

If the jupyter-lab server in the VM is accessed by the browser under windows, files can be directly downloaded from browser and saved in windows.

# DNNDK Lab - Board

# Resize partition

Please enter pynqz2's Jupyter notebook on Windows first.



2.cd jupyter\_notebooks/

```
root@pynq:/home/xilinx# cd jupyter_notebooks/
```

3. ./resize\_partition.sh

```
root@pynq:/home/xilinx/jupyter_notebooks# ./resize_partition.sh
```

You will see:

```
root@pynq:/home/xilinx/jupyter_notebooks# ./resize_partition.sh
Checking initial partition info.....
Filesystem      1K-blocks   Used   Available  Use% Mounted on
/dev/root        7226432  6805216   92772  99% /
devtmpfs          123440     0    123440  0% /dev
tmpfs            255024     0    255024  0% /dev/shm
tmpfs            255024    1252   253772  1% /run
tmpfs             5120     0     5120  0% /run/lock
tmpfs            255024     0    255024  0% /sys/fs/cgroup
Install resize tools
```

```
Finished resizing....
checking partition info
Filesystem      1K-blocks   Used   Available  Use% Mounted on
/dev/root        15064244  6824044   7617064  48%
devtmpfs          123440     0    123440  0% /dev
tmpfs            255024     0    255024  0% /dev/shm
tmpfs            255024    1252   253772  1% /run
tmpfs             5120     0     5120  0% /run/lock
tmpfs            255024     0    255024  0% /sys/fs/cgroup
done!
root@pynq:/home/xilinx/jupyter_notebooks#
```

# Upload mnist.elf and run test notebook

Select items to perform actions on them.

1.Enter pynq-z2\_DL\_project

**pynq-z2\_DL\_project**

Untitled1.ipynb

Welcome to Pynq.ipynb

Upload New

Name Last Modified

0 base common

2 years ago 2 months ago

2.Upload mnist.elf

Select items to perform actions on them.

Upload

Upload New

Name Last Modified

0 .. / pynq-z2\_DL\_project

seconds ago 19 days ago 26 minutes ago 5 minutes ago 19 days ago 19 days ago 19 days ago 3 minutes ago 19 days ago

..

mnist\_gt

mnist\_test-Accuracy.ipynb

mnist\_test.ipynb

resnet50v1\_test-Copy1.ipynb

compile\_model.sh

dpu\_mnist.elf

mnist\_gt.txt

mnist\_test.py

test.py

Select items to perform actions on them.

3.Run mnist\_test.ipynb

mnist\_test.ipynb

resnet50v1\_test-Copy1.ipynb

compile\_model.sh

dpu\_mnist.elf

mnist\_gt.txt

mnist\_test.py

test.py

Upload New

Name Last Modified

0 .. / pynq-z2\_DL\_project

seconds ago 19 days ago 19 days ago 19 days ago 19 days ago 3 minutes ago 19 days ago

# mnist\_test.ipynb

```
import sys
sys.path.append('/usr/local/lib/python2.7/dist-packages')
from dnndk import n2cube, dputils
from ctypes import *
import cv2
import numpy as np
import os
import time
from matplotlib import pyplot as plt

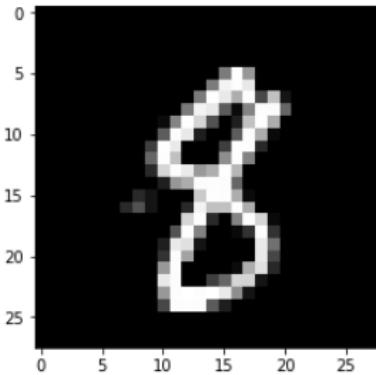
def softmax(x):
    """Compute softmax values for each sets of scores in x"""
    return np.exp(x) / np.sum(np.exp(x), axis=0)

#compile model from elf to .so
!./compile_model.sh

'libdpumodelmnist.so' -> '/usr/lib/libdpumodelmnist.so'

#Load Images
img_path = './mnist_gt'
img = cv2.imread(os.path.join(img_path,'5413.png'))
plt.imshow(img,cmap='gray')

#Pre-processing
img=cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
img = img / 255.0
```



```
KERNEL_CONV = "mnist".encode('utf8')
KERNEL_CONV_INPUT = "conv2d_Conv2D".encode('utf8')
KERNEL_FC_OUTPUT = "dense_1_MatMul".encode('utf8')
# Attach to DPU driver and prepare for running
n2cube.dpuOpen()
# Create DPU Kernels for ResNet50
kernel = n2cube.dpuLoadKernel(KERNEL_CONV)
# Create DPU Tasks from DPU Kernel
task = n2cube.dpuCreateTask(kernel, 0)
# Get the input tensor size from dpu
size = n2cube.dpuGetInputTensorSize(task, KERNEL_CONV_INPUT)
# Get the output tensor channel count from FC output
channel = n2cube.dpuGetOutputTensorSize(task, KERNEL_FC_OUTPUT)

FCResult = [0] * channel #channel=10

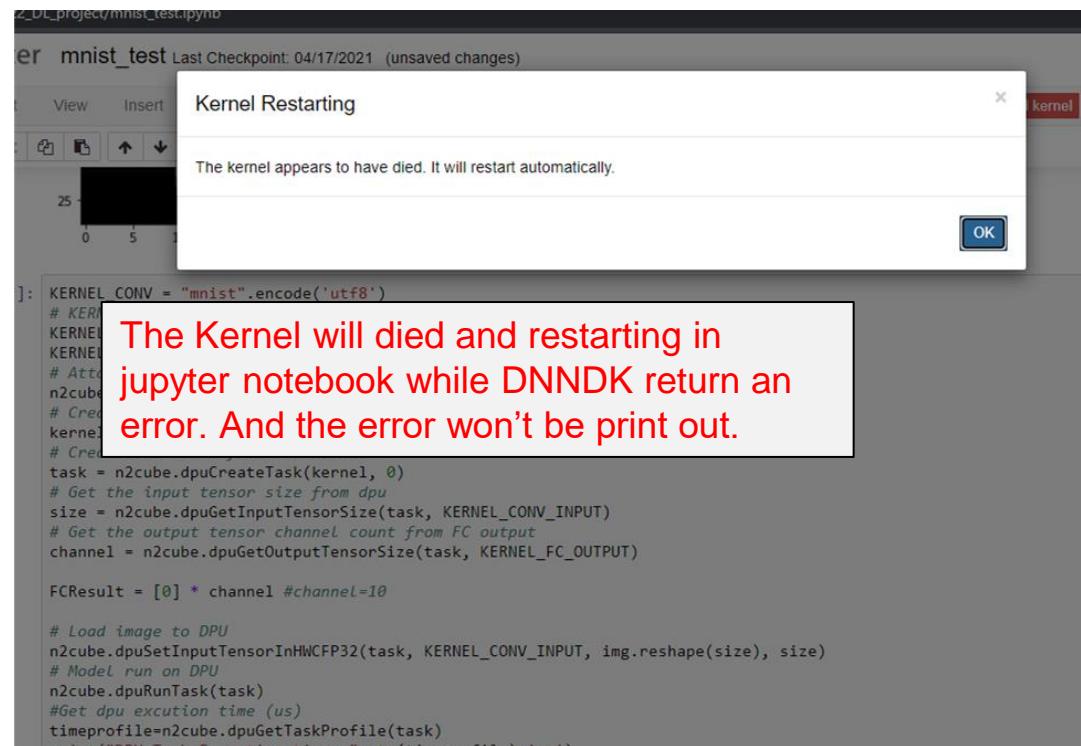
# Load image to DPU
n2cube.dpuSetInputTensorInHWCFP32(task, KERNEL_CONV_INPUT, img.reshape(size), size)
# Model run on DPU
n2cube.dpuRunTask(task)
#Get dpu excution time (us)
timeprofile=n2cube.dpuGetTaskProfile(task)
print("DPU Task Execution time: "+str(timeprofile)+"us")
# Get the output from FC output
n2cube.dpuGetOutputTensorInHWCFP32(task, KERNEL_FC_OUTPUT, FCResult, channel)
#Print Raw dpu output
print("Dpu output: "+str(FCResult))
#caculate softmax
print("Softmax output: ", softmax(FCResult).round(3))
# Get the label
label = FCResult.index(max(FCResult))
print("predict answer: "+str(label))

#close DPU
n2cube.dpuDestroyTask(task)
n2cube.dpuDestroyKernel(kernel)
n2cube.dpuClose()

DPU Task Execution time: 603us
Dpu output: [-3.875, -2.375, -0.5, 3.375, 0.75, 0.0, -4.0, -3.5, 8.0, 1.375]
Softmax output: [ 0.        0.        0.        0.01     0.001   0.        0.        0.        0.988   0.001]
predict answer: 8
```

# Note: About debugging

DNNDK n2cube is hard to debug in jupyter notebook since the notebook is unable to show error code from n2cube. It's recommended to write the code in .py file and execute in command line.



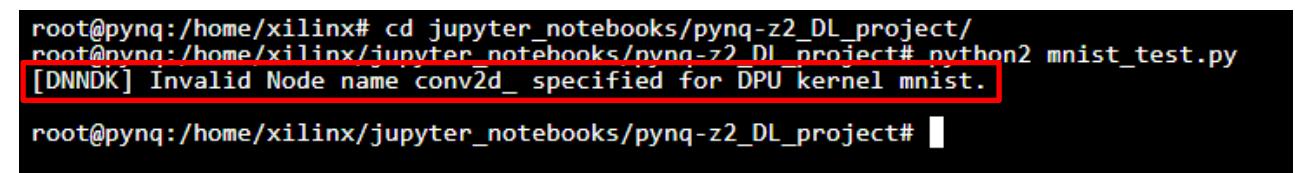
The Kernel will died and restarting in jupyter notebook while DNNDK return an error. And the error won't be print out.

```
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# Create a DNN task
task = n2cube.dpuCreateTask(kernel, 0)
# Get the input tensor size from dpu
size = n2cube.dpuGetInputTensorSize(task, KERNEL_CONV_INPUT)
# Get the output tensor channel count from FC output
channel = n2cube.dpuGetOutputTensorSize(task, KERNEL_FC_OUTPUT)

FCResult = [0] * channel #channel=10

# Load image to DPU
n2cube.dpuSetInputTensorInHWCFP32(task, KERNEL_CONV_INPUT, img.reshape(size), size)
# Model run on DPU
n2cube.dpuRunTask(task)
#Get dpu excution time (us)
timeprofile=n2cube.dpuGetTaskProfile(task)
print("Time profile for task %d: %f us" % (task, timeprofile[0]))
```



Under command line, the error return from DNNDK will be print out and help you debugging.

# Lab goal

- Please trying to modify the code and get the result of model accuracy, FPS
- Please calculate the FPS in entire flow, including pre-processing and post-processing

Ex:

```
accuracy=97.95%
Inference Time=47.882927656173706s, FPS:208.84270217990036
```

Hint: DPU Task can be reuse

Ground truth is in mnist\_gt.txt

Image file path

```
./mnist_gt/0000.png:7
./mnist_gt/0001.png:2
./mnist_gt/0002.png:1
./mnist_gt/0003.png:0
./mnist_gt/0004.png:4
./mnist_gt/0005.png:1
./mnist_gt/0006.png:4
./mnist_gt/0007.png:9
./mnist_gt/0008.png:5
```

answer

Mnist ground truth

Load ground truth reference

```
#Load Image list & gt
gt_path='./mnist_gt.txt'
with open(gt_path,'r') as f:
    gt_list,img_list=[], []
    for line in f.readlines():
        img_list.append(line.split(':')[0])
        gt_list.append(int(line.split(':')[1].replace('\n','')))
```

FPS calculate reference

```
import time
total_image_counts = len(img_list) # 10000 test images in mnist data set
start=time.time()
'''Your code here'''
end=time.time()
fps= total_image_counts / (end-start)
```