Model Compiler Suite for Aries

Developers Guide

version 0.8.1

December 8, 2023



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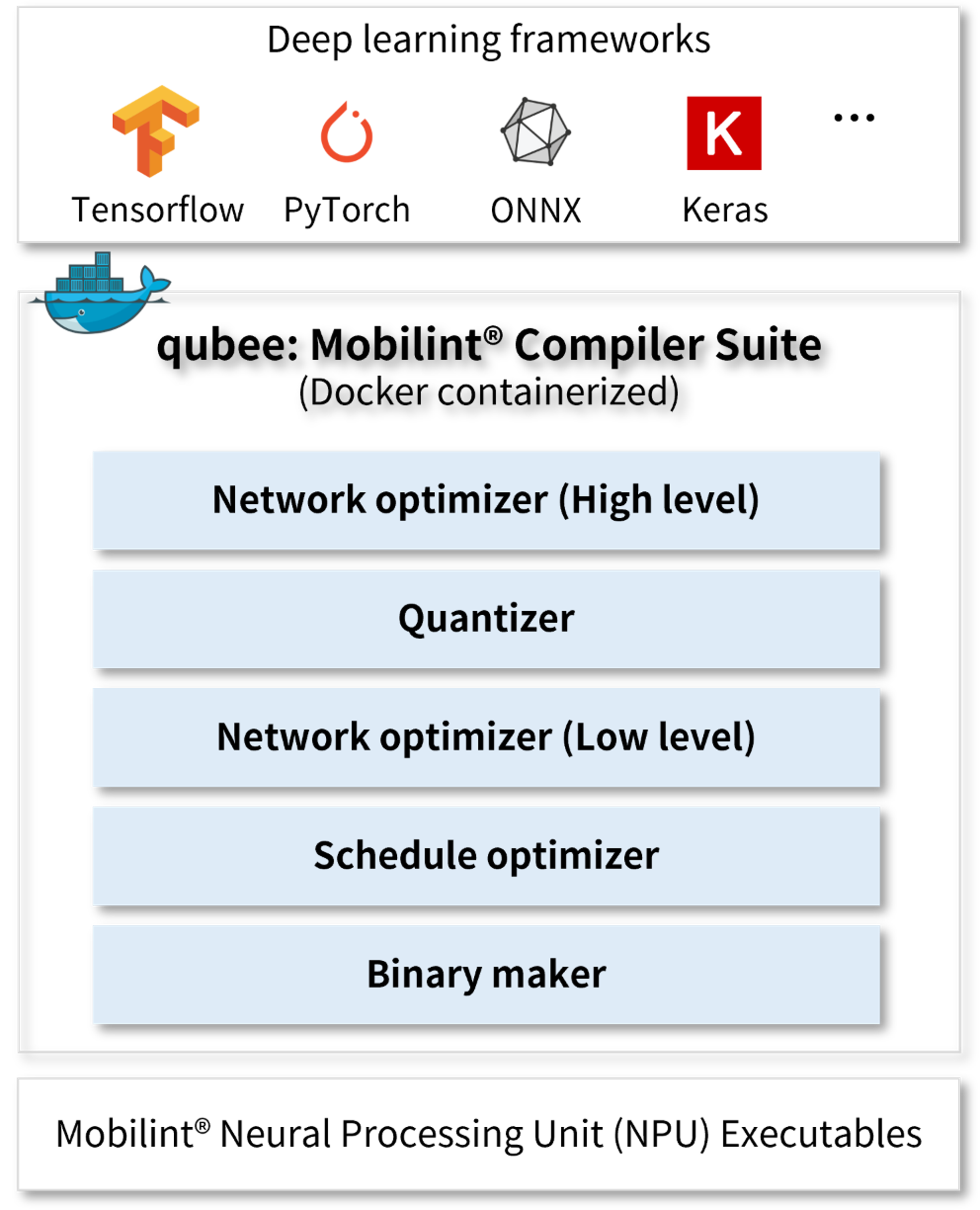
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# Introduction

Mobilint® Model Compiler (i.e., Compiler) is a tool that converts models from deep learning frameworks (ONNX, PyTorch, Keras, TensorFlow, etc...) into Mobilint® Model eXeCUtable (i.e., MXQ), a format executable by Mobilint® Neural Processing Unit (NPU). This is the manual for the **qubee**, Mobilint's SDK. In this manual, you can learn how to use it, what frameworks it supports, etc. A set of functions you can use to interact with the SDK will be given below.



1. Figure 1‑1. SDK Components

Inputs to qubee are a trained deep learning model, its input shape, and calibration data. It will return MXQ (compiled model) as an output.



1. Figure 1‑2. Input and output of qubee

# Changelog

## qubee v0.8.1 (December 2023)

## qubee v0.8.0 (November 2023)

## qubee v0.7.12 (September 2023)

## qubee v0.7.11 (August 2023)

API

Torchscript Backend

## qubee v0.7.10 (August 2023)

## qubee v0.7.9 (August 2023)

## qubee v0.7.8 (August 2023)

## qubee v0.7.7 (June 2023)

API

CPU offloading (beta version)

Improve CPU efficiency

Support more operations

Docker

torch: 1.10.1 -> 1.13.0

tensorflow: 2.3.0 -> 2.9.0

onnx:1.11.0 -> 1.12.0

## qubee v0.7 (March 2023)

Multi-channel quantization

Support more operations

API

Calibration dataset

CPU Offloading (Beta Version)

## qubee v0.6 (August 2022)

Minor updates

## qubee v0.5 (July 2022)

Docker

Conda -> Virtualenv

Python: 3.7.7 -> 3.8.10

torch: 1.8.1 -> 1.10.1

tensorflow: 1.15.0 -> 2.3.0

onnx:1.6.0 -> 1.11.0

Parser

Code refactoring

API

Enable saving sample inference results (inputs and outputs)

## qubee v0.4 (February 2022)

Optimizer

Minor updates in fusing reshape

## qubee v0.3 (February 2022)

Parser

Identify preprocess and postprocess of the model

Exclude preprocess and postprocess if they are unsupported by the NPU

API

Simulate integer inference in Python API

## qubee v0.2 (December 2021)

First release

# Installation

## System requirements

We recommend to use NVIDIA GPU for faster compile wtih qubee, but it is not necessary. Currently, CPU version qubee is also supported.

### Reference System

|  |
| --- |
| Ubuntu 20.04.4 LTS  NVIDIA Graphics Driver 525.147.05 |

### Recommended Packages

|  |
| --- |
| NVIDIA Graphics Driver 450.80.02 **or** Above  Docker  nvidia-docker |

## SDK Installation

We recommend installing qubee on the mobilint docker container.

(Docker image: mobilint/qbcompiler:v0.8.1, <https://hub.docker.com/r/mobilint/qbcompiler>)

### Building Docker Image

Run the following commands to build the docker image.

|  |
| --- |
| $ *# Docker image download*  $ docker pull mobilint/qbcompiler:v0.8.1  $ *# Make a docker container (if needed)*  $ *# mkdir {WORKING DIRCTORY}*  $ cd {WORKING DIRCTORY}  $ docker run -it --gpus all --ipc=host --name mxq\_compiler -v $(pwd):/workspace mobilint/qbcompiler:v0.8.1 |

(Recommended) If the trained models and calibration dataset are stored in different directories, you can mount them to the docker container as follows:

|  |
| --- |
| $ docker run -it --gpus all --ipc=host --name mxq\_compiler -v $(pwd):/workspace -v {PATH TO MODEL DIR}:/models -v {PATH TO CALIBRATION DATASET DIR}:/datasets mobilint/qbcompiler:v0.8.1 |

(Option, not available yet) Build the docker image for WSL2

|  |
| --- |
| $ *# Docker image download*  $ docker pull mobilint/qbcompiler:v0.8.1-wsl  $ *# Make a docker container*  $ cd {WORKING DIRCTORY}  $ docker run -it --gpus all --ipc=host --name mxq\_compiler -v $(pwd):/data mobilint/qbcompiler:v0.8.1-wsl |

### Installation of qubee

qubee compiler packages are available in [Mobilint® Software Development Kit (SDK)](https://dl.mobilint.com/view.php%20)

Run the following commands to install qubee on the docker container.

|  |
| --- |
| $ *# Download qubee-0.8.1-py3-none-any.whl file*  $ *# Copy qubee whl file to Docker*  $ docker cp {Path to qubee-0.8.1-py3-none-any.whl} mxq\_compiler:/  $ *# Start docker*  $ docker start mxq\_compiler  $ *# Attach docker*  $ docker **exec** -it mxq\_compiler /bin/bash  $ *# Install qubee*  $ cd /  $ python -m pip install qubee-0.8.1-py3-none-any.whl |

(Option, for WSL2) Run the following commands to install qubee on the docker container.

|  |
| --- |
| $ *# Download qubee-0.8.1\_wsl-py3-none-any.whl file*  $ *# Copy qubee whl file to Docker*  $ docker cp {Path to qubee-0.8.1\_wsl-py3-none-any.whl} mxq\_compiler:/  $ *# Start docker*  $ docker start mxq\_compiler  $ *# Attach docker*  $ docker **exec** -it mxq\_compiler /bin/bash  $ *# Install qubee*  $ cd /  $ python -m pip install qubee-0.8.1\_wsl-py3-none-any.whl |

# Tutorials

The tutorials below go through the steps for preparing the calibration dataset, model compiling, and inference.

## Preparing Calibration Data

To compile the model, you should prepare the calibration dataset (the pre-processed inputs for the model) for quantization. There are three ways to make the calibration dataset as follows:

(i) Pre-process the raw calibration dataset and save it as numpy tensors.

(ii) Utilize a pre-processing configuration YAML file (only for images with **uniform format**).

(iii) Use a manually defined pre-processing function (only for images with **uniform format**).

(iv) Use Mobilint® Processor (will be available soon)

**Important**  The process of making a calibration dataset may vary depending on whether you compile the model for CPU offloading or not. Currently, qubee compiles the model without CPU offloading by default. In this scenario, the pre-processed input shape should be in the format (H, W, C). On the other hand, when CPU offloading is employed, the pre-processed input shape should match the input shape that the original model takes.

### Pre-process raw calibration dataset and save it as numpy tensors

You can save the pre-processed calibration dataset as numpy tensors with your custom pre-processing function and use them to compile the model.

An example code is shown below. The following code assumes that we hold an image folder consisting of 1000 randomly selected images from the Imagenet dataset for calibration prepared in directory `/datasets/imagenet/cali\_1000`.

|  |
| --- |
| **import** os  **import** numpy as np  **import** cv2  **def** **get\_img\_paths\_from\_dir**(dir\_path: str, img\_ext = ["jpg", "jpeg", "png"]):  **assert** os.path.**exists**(dir\_path)  candidates = os.**listdir**(dir\_path)  **return** [os.path.**join**(dir\_path, y) **for** y **in** candidates **if** **any**([y.**lower**().**endswith**(e) **for** e **in**  img\_ext])]  **def** **pre\_process**(img\_path: str, target\_h: int, target\_w: int):  img = cv2.**imread**(img\_path, cv2.IMREAD\_COLOR)  resized\_img = cv2.**resize**(img, dsize=(target\_w, target\_h)).**astype**(np.float32)  **return** resized\_img  **if** \_\_name\_\_ == "\_\_main\_\_":  img\_dir = "/datasets/imagenet/cali\_1000"  save\_dir = "/workspace/calibration/custom\_single\_input"  target\_h, target\_w = 224, 224  os.**makedirs**(save\_dir, exist\_ok=True)  img\_paths = **get\_img\_paths\_from\_dir**(img\_dir)  **for** i, img\_path **in** **enumerate**(img\_paths):  fname = f"{i}".**zfill**(3) + ".npy"  fpath = os.path.**join**(save\_dir, fname)  x = **pre\_process**(img\_path, target\_h, target\_w)  np.**save**(fpath, x) |

The above results in a directory containing the pre-processed calibration dataset (numpy tensors of shape (224,224, 3)), located at `/workspace/calibration/custom\_single\_input`.

### Use a pre-processing configuration YAML file

Image pre-processing techniques such as resizing, cropping, and normalization are often applied in machine vision tasks. Users can construct a pre-processing configuration using a YAML file and prepare the calibration dataset via the API provided by qubee, *make\_calib*. Please be aware that this method can only be employed when the raw data is an image. An example code is shown below. The following code assumes that images for calibration are prepared in the directory `/workspace/cali\_1000`.

|  |
| --- |
| **from** qubee **import** make\_calib  **make\_calib**(  args\_pre="/workspace/mobilenet\_v2.yaml", *# path to pre-processing configuration yaml file*  data\_dir="/datasets/imagenet/cali\_1000", *# path to folder of original calibration data files such as images*  save\_dir="/workspace/calibration/", *# path to folder to save pre-proceessed calibration data files*  save\_name="mobilenet\_v2", *# tag for the generated calibration dataset*  max\_size=50 *# Maximum number of data to use for calibration*  ) |
| *# mobilenet\_v2.yaml*  Datatype: Image  GetImage:  to\_float32: **false**  channel\_order: RGB  Pre-Order: [ResizeTorch, CenterCrop, Normalize, SetOrder]  Pre-processing:  ResizeTorch:  size: [256, 256]  interpolation: bilinear  CenterCrop:  size: [224, 224]  Normalize:  mean: [0.485, 0.456, 0.406]  std: [0.229, 0.224, 0.225]  to\_float\_div255: **true**  SetOrder:  shape: HWC |

The above results are in a directory containing the pre-processed calibration dataset (numpy tensors), located at `/workspace/calibration/mobilenet\_v2`. In addition, a calibration meta txt file containing the paths to the pre-processed numpy files is created, named `/workspace/calibration/mobilenet\_v2.txt`.

**Remark** The sample dataset for calibration should be composed of images with the same format. If some are in color images and others are in grayscale images, the calibration dataset will not be created properly.

### Use a manually defined pre-processing function

You can use your pre-processing function to make the calibration dataset via the API provided by qubee, *make\_calib\_man*. In this case, the pre-processing function should take the image path as input and return a numpy tensor. An example of the code is shown below. The following code assumes that images for calibration are prepared in the directory `/datasets/imagenet/cali\_1000`.

|  |
| --- |
| **import** torch  **import** numpy as np  **from** PIL **import** Image  **import** torchvision.transforms.functional as F  **from** qubee **import** make\_calib\_man  **def** **preprocess\_resnet50**(img\_path: str):  img = Image.**open**(img\_path)  resize\_size=(232, 232)  crop\_size=(224, 224)  mean=[0.485, 0.456, 0.406]  std=[0.229, 0.224, 0.225]  out = F.**pil\_to\_tensor**(img)  out = F.**resize**(out, size=resize\_size)  out = F.**center\_crop**(out, output\_size=crop\_size)  out = out.**to**(torch.float, copy=False) / 255.  out = F.**normalize**(out, mean, std)  out = np.**transpose**(out.**numpy**(), axes=[1, 2, 0])  **return** out  **make\_calib\_man**(  pre\_ftn=preprocess\_resnet50, *# callable function to pre-process the calibration data*  data\_dir="/datasets/imagenet/cali\_1000", *# path to folder of original calibration data files such as images*  save\_dir="/workspace/calibration/", *# path to folder to save pre-proceessed calibration data files*  save\_name="resnet50", *# tag for the generated calibration dataset*  max\_size=50 *# Maximum number of data to use for calibration*  ) |

The above results are in a directory containing the pre-processed calibration dataset (numpy tensors), located at `/workspace/sample/calibration/resnet50`. In addition, a calibration meta txt file containing the paths to the pre-processed numpy files is created, named `/workspace/sample/calibration/resnet50.txt`.

**Remark** Unless the custom pre-processing function contains proper exception handling, the sample dataset for calibration should be composed of images with the same format. Like the previous method, the calibration dataset will not be created properly if some are in color images and others are in grayscale images.

## Compiling ONNX Models

ONNX model is recommended to use for compiling the trained model. With simple code, the ONNX model can be directly parsed to obtain Mobilint IR. example code is shown below. The following code assumes that the calibration dataset and the model are prepared in the directory `/workspace/calibration/resnet50` and `/workspace/resnet50.onnx`, respectively.

|  |
| --- |
| *""" Compile ONNX model"""*  **from** qubee **import** mxq\_compile  onnx\_model\_path = "/workspace/resnet50.onnx"  calib\_data\_path = "/workspace/calibration/resnet50"  *# calib\_data\_path can be replaced with the path to the calibration meta file such as "/workspace/calibration/resnet50.txt"*  **mxq\_compile**(  model=onnx\_model\_path,  calib\_data\_path=calib\_data\_path,  save\_path="resnet50.mxq",  backend="onnx"  ) |

## Compiling PyTorch Models

PyTorch models can be compiled in two different ways. The first approach involves converting the PyTorch model to ONNX, which is then further converted into Mobilint IR. The second approach involves converting the PyTorch model to Torchscript, which is then further converted into Mobilint IR. Once the model is converted to Mobilint IR, then it is be compiled into MXQ. Examples of the code are shown below. The following codes assume that the calibration dataset is prepared in directory `/workspace/calibration/resnet50`.

|  |
| --- |
| *""" Compile PyTorch model, first way """*  **from** qubee **import** mxq\_compile  **from** qubee.utils **import** convert\_pytorch\_to\_onnx  **import** torchvision  input\_shape = (224, 224, 3)  calib\_data\_path = "/workspace/calibration/resnet50"  *# A calibration meta file such as "/workspace/calibration/resnet50.txt" can be used instead.*  *### get resnet50 from torchvision and convert it to ONNX*  torch\_model = torchvision.models.**resnet50**(pretrained=True)  onnx\_model\_path = "/workspace/resnet50.onnx"  **convert\_pytorch\_to\_onnx**(torch\_model, input\_shape, onnx\_model\_path)  **mxq\_compile**(  model=onnx\_model\_path,  calib\_data\_path=calib\_data\_path,  save\_path="resnet50.mxq",  backend="onnx"  ) |
| *""" Compile PyTorch model, second way """*  **from** qubee **import** mxq\_compile  *### get resnet50 from torchvision*  **import** torchvision  **import** torch  calib\_data\_path = "/workspace/calibration/resnet50"  *# A calibration meta file such as "/workspace/calibration/resnet50.txt" can be used instead.*  *### get resnet50 from torchvision and convert it to torchscript*  torch\_model = torchvision.models.**resnet50**(pretrained=True)  torchscript\_model\_path = "/workspace/resnet50.pt"  example\_input = torch.**rand**(1, 3, 224, 224)  scripted\_model = torch.jit.**script**(torch\_model, example\_input)  torch.jit.**save**(scripted\_model, torchscript\_model\_path)  **mxq\_compile**(  model=torchscript\_model\_path,  calib\_data\_path=calib\_data\_path,  backend="torchscript",  save\_path="resnet50.mxq",  example\_input=example\_input  ) |

## Compiling Keras/TensorFlow Models

Since Keras works as an interface for TensorFlow 2, models on the Keras framework can be converted to Mobilint IR via TensorFlow. First, we load and save the Keras/TensorFlow model into the format of the frozen graph, which ends with `.pb`. Then, with the directory containing the frozen graph, qubee will compile the model. The following code assumes the calibration dataset is prepared in the directory `/workspace/calibration/resnet50`.

**Remark** According to the annotations and old version instructions, the TensorFlow compilation should work by providing the directory containing the frozen graph or just the frozen graph file. However, the current version makes various errors, such as kernel parsing errors, incompatible tag errors, etc. We are currently working on this issue.

|  |
| --- |
| *""" Compile Keras/TensorFlow model """*  **from** qubee **import** mxq\_compile  **import** tensorflow as tf  keras\_model = tf.keras.applications.resnet50.**ResNet50**() *# Load a Keras model*  input\_shape = (224, 224, 3)  calib\_data\_path = "/workspace/calibration/resnet50"  *# A calibration meta file such as "/workspace/calibration/resnet50.txt" can be used instead.*  keras\_model\_save\_path = "/workspace/tf\_models/resnet50" *# directory to save the model*  keras\_model.**save**(keras\_model\_save\_path) *# Save the model in the format of frozen graph. saved\_model.pb file will be created in the directory.*  keras\_model.**summary**() *# if you are not aware of the input name, you can check it by this command.*  **mxq\_compile**(  model=keras\_model\_save\_path,  calib\_data\_path=calib\_data\_path,  backend="tf1", *# or "tf2". It will be unified to "tf" in the future.*  save\_path="resnet50.mxq",  input\_shape={'input\_1':(224, 224, 3)} *# dictionary of input shape*  ) |

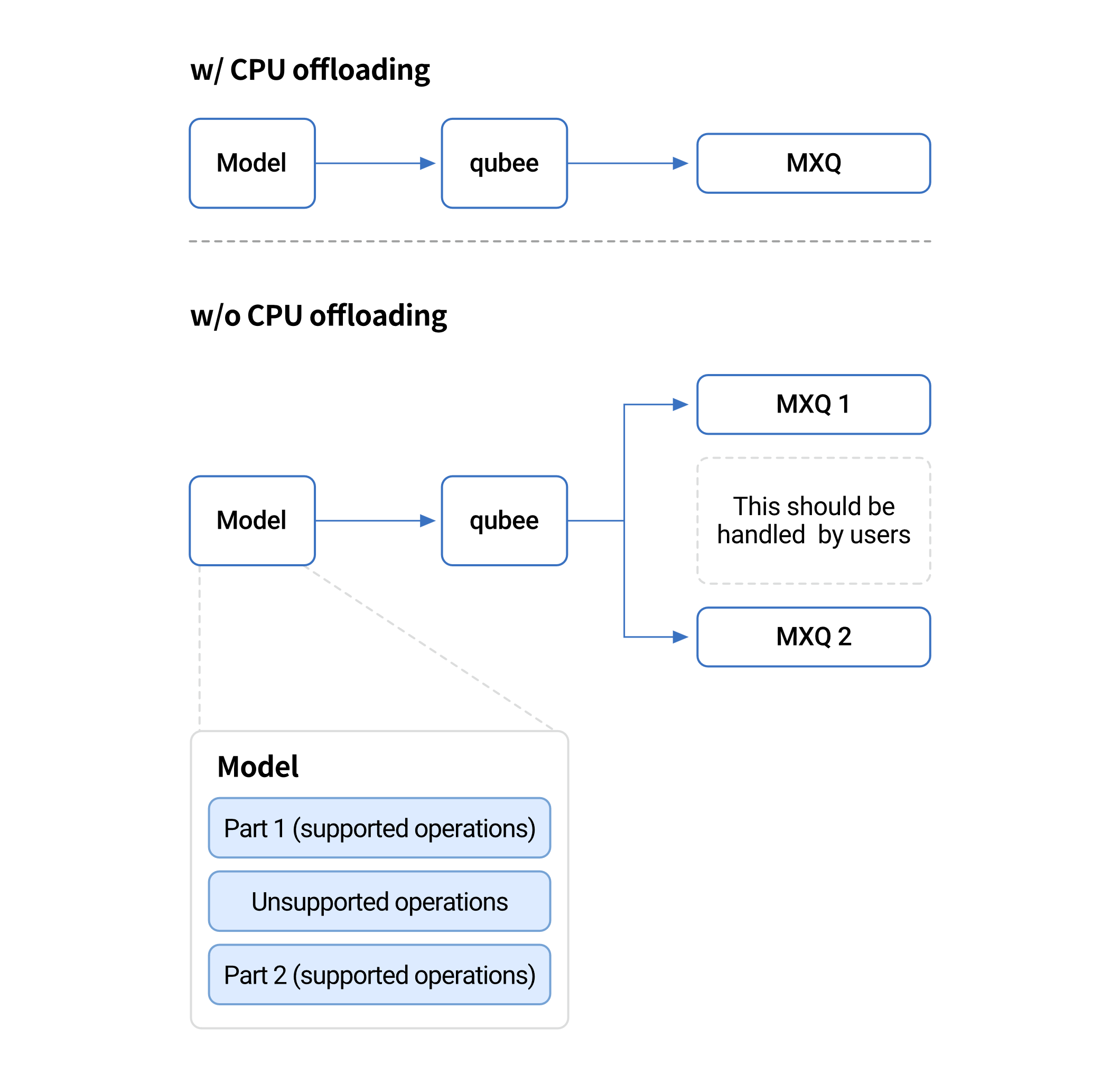
# CPU Offloading

From qubee v0.7, we provide a Beta version of CPU offloading for mxq compile. CPU offloading makes it easier for users to compile their models by automatically offloading the computation that Mobilint NPU does not support to the CPU. For example, if a pre-processing or post-processing included in the model involves operations that the NPU does not support, the user would have to implement them manually after compiling, but CPU offloading covers most of these operations and eliminates the need for additional work.

When CPU offloading is employed, the procedures for preparing the calibration dataset and compiling the model vary slightly as follows:

(i) The pre-processed input shape should match the original model's input shape, whereas the pre-processed input shape should be in the format (H, W, C) to compile the model without CPU offloading.

(ii) Set the argument *cpu\_offload* of function *mxq\_compile* True to enable CPU offloading.



1. Figure 5‑1. SDK CPU Offloading

# Supported Frameworks

We support almost all the commonly used Machine Learning frameworks & libraries such as ONNX, PyTorch, Keras, and TensorFlow.



1. Figure 6‑1. Supported deep-learning frameworks

## Supported Operations (ONNX)

Table 6‑1. ONNX Supported Operations

| API Name | Comments |
| --- | --- |
| [Add](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Add) | Broadcast only for specific cases of constant addition:  Adding scalar,  Adding channel-size vector. |
| [And](https://github.com/onnx/onnx/blob/main/docs/Operators.md#And) |  |
| [ArgMax](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ArgMax) |  |
| [AveragePool](https://github.com/onnx/onnx/blob/main/docs/Operators.md#AveragePool) | Only dilation=1, count\_include\_pad=1. |
| [BatchNormalization](https://github.com/onnx/onnx/blob/main/docs/Operators.md#BatchNormalization) | Only training\_mode=0. |
| [Cast](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Cast) |  |
| [Ceil](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Ceil) |  |
| [Clip](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Clip) |  |
| [Concat](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Concat) | Only along channel axis. |
| [Constant](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Constant) |  |
| [ConstantOfShape](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ConstantOfShape) |  |
| [Conv](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Conv) |  |
| [ConvTranspose](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ConvTranspose) |  |
| [DepthToSpace](https://github.com/onnx/onnx/blob/main/docs/Operators.md#DepthToSpace) |  |
| [Div](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Div) | Only constant division.  Support broadcast same as Add. |
| [Elu](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Elu) |  |
| [Equal](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Equal) |  |
| [Erf](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Erf) |  |
| [Exp](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Exp) |  |
| [Expand](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Expand) |  |
| [Flatten](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Flatten) | Only axis=1 and before fully connected layer or Conv w/ 1x1 kernel. |
| [Floor](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Floor) |  |
| [Gather](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Gather) |  |
| [GatherND](https://github.com/onnx/onnx/blob/main/docs/Operators.md#GatherND) |  |
| [Gemm](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Gemm) | Only transA=0.  Only for the following specific case:  Input A is a flatten activation and input B is 2D tensor. |
| [GlobalAveragePool](https://github.com/onnx/onnx/blob/main/docs/Operators.md#GlobalAveragePool) |  |
| [Greater](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Greater) |  |
| [HardSigmoid](https://github.com/onnx/onnx/blob/main/docs/Operators.md#HardSigmoid) |  |
| [HardSwish](https://github.com/onnx/onnx/blob/main/docs/Operators.md#HardSwish) |  |
| [Identity](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Identity) |  |
| [InstanceNormalization](https://github.com/onnx/onnx/blob/main/docs/Operators.md#InstanceNormalization) |  |
| [LayerNormalization](https://github.com/onnx/onnx/blob/main/docs/Operators.md#LayerNormalization) |  |
| [LeakyRelu](https://github.com/onnx/onnx/blob/main/docs/Operators.md#LeakyRelu) |  |
| [Less](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Less) |  |
| [Loop](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Loop) |  |
| [MatMul](https://github.com/onnx/onnx/blob/main/docs/Operators.md#MatMul) | Only for the following specific case:  Input A is a flatten activation and input B is 2D tensor or vice-versa. |
| [Max](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Max) |  |
| [MaxPool](https://github.com/onnx/onnx/blob/main/docs/Operators.md#MaxPool) | Only dilation=1. |
| [Min](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Min) |  |
| [Mod](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Mod) |  |
| [Mul](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Mul) | Only constant multiplication.  Support broadcast same as Add. |
| [NonMaxSuppression](https://github.com/onnx/onnx/blob/main/docs/Operators.md#NonMaxSuppression) |  |
| [NonZero](https://github.com/onnx/onnx/blob/main/docs/Operators.md#NonZero) |  |
| [Not](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Not) |  |
| [Or](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Or) |  |
| [Pad](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Pad) | Constant, reflect, edge modes are supported |
| [Pow](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Pow) |  |
| [PRelu](https://github.com/onnx/onnx/blob/main/docs/Operators.md#PRelu) |  |
| [Range](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Range) |  |
| [Reciprocal](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Reciprocal) |  |
| [ReduceMax](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ReduceMax) | Only along height and width. |
| [ReduceMean](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ReduceMean) | Only along height and width. |
| [ReduceMin](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ReduceMin) | Only along height and width. |
| [ReduceProd](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ReduceProd) | Only along height and width. |
| [ReduceSum](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ReduceSum) | Only along height and width. |
| [Relu](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Relu) | Only scalar slope. |
| [Reshape](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Reshape) | Only channel-wise flatten and before fully connected layer or Conv w/ 1x1 kernel.  Only allowzero=0. |
| [Resize](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Resize) | Only for the following specific case:  Only mode = "nearest" and coordinate\_transformation\_mode = "half\_pixel" or "pytorch\_half\_pixel",  Only mode = "linear" and coordinate\_transformation\_mode = "half\_pixel" or "pytorch\_half\_pixel",  Attributes axes, antialias, keep\_aspect\_ratio\_policy nearest\_mode are not supported. |
| [ScatterND](https://github.com/onnx/onnx/blob/main/docs/Operators.md#ScatterND) |  |
| [Shape](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Shape) |  |
| [Sigmoid](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Sigmoid) |  |
| [Slice](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Slice) | Only channel-wise slice. |
| [Softmax](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Softmax) |  |
| [Softplus](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Softplus) |  |
| [Split](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Split) |  |
| [Sqrt](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Sqrt) |  |
| [Squeeze](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Squeeze) | Only when resulting tensor has 2D shape.  Squeeze along batch axis is unsupported. |
| [Sub](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Sub) | Support broadcast same as Add. |
| [Tanh](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Tanh) |  |
| [Tile](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Tile) | Batch-wise tile is unsupported. |
| [TopK](https://github.com/onnx/onnx/blob/main/docs/Operators.md#TopK) |  |
| [Transpose](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Transpose) | Only for the following specific case:  Transpose-Flatten-Linear. |
| [Unsqueeze](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Unsqueeze) |  |
| [Upsample](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Upsample) | Only mode "nearest" and "linear". |
| [Xor](https://github.com/onnx/onnx/blob/main/docs/Operators.md#Xor) |  |

## Supported operations (PyTorch)

**Remark** Since the Torchscript backend framework is based on [Torchscript-Based-ONNX-Exporter](https://pytorch.org/docs/stable/onnx.html#Torchscript-Based-ONNX-Exporter), even if the operation is not listed below, it can be supported if it has corresponding ONNX operation.

Table 6‑2. PyTorch Supported Operations

| API Name | Comments |
| --- | --- |
| [nn.Conv2d](https://pytorch.org/docs/1.13/generated/torch.nn.Conv2d.html#torch.nn.Conv2d) |  |
| [nn.ConvTranspose2d](https://pytorch.org/docs/1.13/generated/torch.nn.ConvTranspose2d.html#torch.nn.ConvTranspose2d) |  |
| [nn.Linear](https://pytorch.org/docs/1.13/generated/torch.nn.Linear.html#torch.nn.Linear) |  |
| [nn.BatchNorm2d](https://pytorch.org/docs/1.13/generated/torch.nn.BatchNorm2d.html#torch.nn.BatchNorm2d) |  |
| [nn.MaxPool2d](https://pytorch.org/docs/1.13/generated/torch.nn.MaxPool2d.html#torch.nn.MaxPool2d) | Only dilation=1. |
| [nn.AvgPool2d](https://pytorch.org/docs/1.13/generated/torch.nn.AvgPool2d.html#torch.nn.AvgPool2d) | Only dilation=1, count\_include\_pad=1. |
| [nn.AdaptiveAvgPool2d](https://pytorch.org/docs/1.13/generated/torch.nn.AdaptiveAvgPool2d.html#torch.nn.AdaptiveAvgPool2d) |  |
| [nn.functional.interpolate](https://pytorch.org/docs/1.13/generated/torch.nn.functional.interpolate.html#torch.nn.functional.interpolate) | See supported operations (ONNX): resize. |
| [nn.Upsample](https://pytorch.org/docs/1.13/generated/torch.nn.functional.upsample.html#torch.nn.functional.upsample) | Only mode "nearest" and "linear".  Only scales=[2,2]. |
| [Add](https://pytorch.org/docs/1.13/generated/torch.add.html#torch.add) | Only alpha=1. |
| [Sub](https://pytorch.org/docs/1.13/generated/torch.sub.html#torch.sub) | Only alpha=1. |
| [Mul](https://pytorch.org/docs/1.13/generated/torch.mul.html#torch.mull) | Only constant multiplication. |
| [Div](https://pytorch.org/docs/1.13/generated/torch.div.html#torch.div) | Only constant division. |
| [Cat](https://pytorch.org/docs/1.13/generated/torch.cat.html#torch.cat) | Only along channel axis. |
| [Relu](https://pytorch.org/docs/1.13/generated/torch.nn.ReLU.html#torch.nn.ReLU) |  |
| [PRelu](https://pytorch.org/docs/1.13/generated/torch.nn.PReLU.html#torch.nn.PReLU) |  |
| [LeakyRelu](https://pytorch.org/docs/1.13/generated/torch.nn.LeakyReLU.html#torch.nn.LeakyReLU) |  |
| [Sigmoid](https://pytorch.org/docs/1.13/generated/torch.nn.Sigmoid.html#torch.nn.Sigmoid) |  |
| [Tanh](https://pytorch.org/docs/1.13/generated/torch.nn.Tanh.html#torch.nn.Tanh) |  |
| [Softplus](https://pytorch.org/docs/1.13/generated/torch.nn.Softplus.html#torch.nn.Softplus) | Only beta=1. |
| [Hardswish](https://pytorch.org/docs/1.13/generated/torch.nn.Hardswish.html#torch.nn.Hardswish) |  |
| [Clip](https://pytorch.org/docs/1.13/generated/torch.clip.html#torch.clip) |  |
| [Exp](https://pytorch.org/docs/1.13/generated/torch.exp.html#torch.exp) |  |
| [Reshape](https://pytorch.org/docs/1.13/generated/torch.reshape.html#torch.reshape) | Only channel-wise flatten and before fully connected layer or Conv w/ 1x1 kernel. |
| [Transpose](https://pytorch.org/docs/1.13/generated/torch.transpose.html#torch.transpose) | Only before fully connected layer. |
| [Squeeze](https://pytorch.org/docs/1.13/generated/torch.squeeze.html#torch.squeezev) | Only when resulting tensor has 2D shape.  Squeeze along batch axis is unsupported. |
| [nn.Flatten](https://pytorch.org/docs/1.13/generated/torch.nn.Flatten.html#torch.nn.Flatten) | Only channel-wise flatten and before fully connected layer or Conv w/ 1x1 kernel. |
| [nn.Identity](https://pytorch.org/docs/1.13/generated/torch.nn.Identity.html#torch.nn.Identity) |  |
| [Pad](https://pytorch.org/docs/1.13/generated/torch.nn.functional.pad.html#torch.nn.functional.pad) |  |
| [Ceil](https://pytorch.org/docs/1.13/generated/torch.ceil.html#torch.ceil) |  |
| [Clamp](https://pytorch.org/docs/1.13/generated/torch.clamp.html#torch.clamp) | See supported operations (ONNX): clip |
| [Erf](https://pytorch.org/docs/1.13/special.html#torch.special.erf) |  |
| [Exp](https://pytorch.org/docs/1.13/generated/torch.exp.html#torch.exp) |  |
| [Floor](https://pytorch.org/docs/1.13/generated/torch.floor.html#torch.floor) |  |
| [Pow](https://pytorch.org/docs/1.13/generated/torch.pow.html#torch.pow) |  |
| [Reciprocal](https://pytorch.org/docs/1.13/generated/torch.reciprocal.html#torch.reciprocal) |  |

## Supported operations (TensorFlow)

Table 6‑3. TensorFlow Supported Operations

| API Name | Comments |
| --- | --- |
| [tf.placeholder](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/placeholder) |  |
| [tf.constant](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/constant) |  |
| [tf.identity](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/identity) |  |
| [tf.identity\_n](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/identity_n) |  |
| [tf.pad](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/pad) |  |
| [tf.nn.conv2d](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/conv2d) |  |
| [tf.nn.depthwise\_conv2d](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/depthwise_conv2d) |  |
| [tf.nn.conv2d\_backprop\_input](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/conv2d_backprop_input) |  |
| [tf.linalg.matmul](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/linalg/matmul) |  |
| [tf.nn.fused\_batch\_norm](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/fused_batch_norm) |  |
| [tf.nn.max\_pool2d](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/max_pool2d) |  |
| [tf.nn.avg\_pool](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/avg_pool) |  |
| [tf.math.reduce\_mean](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/reduce_mean) | Only along height, width, and channel. |
| [tf.image.resize\_nearest\_neighbor](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/image/resize_nearest_neighbor) |  |
| tf.raw\_ops.ConcatV2 | Only along channel axis. |
| [tf.math.add\_n](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/add_n) |  |
| [tf.nn.bias\_add](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/bias_add) |  |
| [tf.math.multiply](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/multiply) | Only constant multiplication. |
| [tf.realdiv](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/realdiv) | Only constant division. |
| [tf.math.subtract](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/subtract) |  |
| [tf.nn.relu](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/relu) |  |
| [tf.nn.leaky\_relu](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/leaky_relu) |  |
| [tf.nn.relu6v](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/nn/relu6#args) |  |
| [tf.math.sigmoid](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/sigmoid) |  |
| [tf.math.softplus](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/softplus) |  |
| [tf.math.tanh](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/tanh) | Only beta=1. |
| [tf.math.exp](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/math/exp) |  |
| [tf.switch\_case](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/switch_case) |  |
| [tf.shape](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/shape) |  |
| [tf.reshape](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/reshape) | Only channel-wise flatten and before fully connected layer or Conv w/ 1x1 kernel. |
| [tf.transpose](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/transpose) | Only before fully connected layer. |
| [tf.expand\_dims](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/expand_dims) |  |
| [tf.squeeze](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/squeeze) | Only when resulting tensor has 2D shape.  Squeeze along batch axis is unsupported. |
| [tf.strided\_slice](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/strided_slice) | ellipsis\_mask, new\_axis\_mask, shrink\_axis\_mask are unsupported. |
| [tf.tile](https://www.tensorflow.org/versions/r1.15/api_docs/python/tf/tile) | Batch-wise tile is unsupported. |

## Supported operations (Keras)

Table 6‑4. Keras Supported Operations

| API Name | Comments |
| --- | --- |
| [layers.Conv2D](https://keras.io/api/layers/convolution_layers/convolution2d/) |  |
| [layers.Conv2DTranspose](https://keras.io/api/layers/convolution_layers/convolution2d_transpose/) |  |
| [layers.Dense](https://keras.io/api/layers/core_layers/dense/) |  |
| [layers.BatchNormalization](https://keras.io/api/layers/normalization_layers/batch_normalization/) | Only training=False. |
| [layers.ZeroPadding2D](https://keras.io/api/layers/reshaping_layers/zero_padding2d/) |  |
| [layers.Concatenate](https://keras.io/api/layers/merging_layers/concatenate/) | Only along channel axis. |
| [layers.Add](https://keras.io/api/layers/merging_layers/add/) |  |
| [layers.Subtract](https://keras.io/api/layers/merging_layers/subtract/) |  |
| [layers.Multiply](https://keras.io/api/layers/merging_layers/multiply/) | Only constant multiplication. |
| [layers.MaxPooling2D](https://keras.io/api/layers/pooling_layers/max_pooling2d/) |  |
| [layers.AveragePooling2D](https://keras.io/api/layers/pooling_layers/average_pooling2d/) |  |
| [layers.GlobalAveragePooling2D](https://keras.io/api/layers/pooling_layers/global_average_pooling2d/) |  |
| [layers.Dropout](https://keras.io/api/layers/regularization_layers/dropout/) | Only inference mode. |
| [layers.Flatten](https://keras.io/api/layers/reshaping_layers/flatten/) | Only channel-wise flatten and before fully connected layer or Conv w/ 1x1 kernel. |
| [activations.relu](https://keras.io/api/layers/activations/) | Only alpha=0, max\_value=None, threshold=0. |
| [activations.sigmoid](https://keras.io/api/layers/activations/) |  |
| [activations.softplus](https://keras.io/api/layers/activations/) |  |
| [activations.tanh](https://keras.io/api/layers/activations/) |  |
| [activations.exponential](https://keras.io/api/layers/activations/) |  |

# API Reference

## Class: Model\_Dict

This class serves two main functions:

1. Compile

2. Inference (Note that this inference is only for testing and done by CPU or GPU.)

Table 7‑1. Model\_Dict Class

| Attributes | Type | Description |
| --- | --- | --- |
| model\_dict | ONNX\_Model\_Dict, TF\_Model\_Dict | Mobilint IR, which holds information of layers in the model. |
| model\_from | string | Backend for holding information of the model. |
| output\_name\_list | List[string] | List of the keys (in model\_dict) corresponding to the output layer of the model. (It could be different from the original model, because qubee parses deep learning related operations only.) |
| model\_from | string | Deep learning framework where the input model comes from. |
| c\_model | qubee.mmc.Compiler | Low-level compiler. (defined in C++ code). It compiles Mobilint IR into MXQ format. |
| p\_model | qubee.model\_dict.Model | Model restored from Mobilint IR. This enables full-precision inference for testing. |
| is\_compiled | bool | This indicates whether the model is compiled. |
| device | string | Device to be used for compile and inerence. Either cpu or gpu. |
| has\_c\_model | bool | This indicates whether the c\_model is prepared. |
| has\_p\_model | bool | This indicates whether the p\_model is prepared. |

### Methods

Table 7‑2. Model\_Dict Methods

| Methods | Description |
| --- | --- |
| \_\_init\_\_ | Constructor of Mobilint IR model. |
| compile | Compile the given model into MXQ format. |
| inference | Floating inference with the Mobilint IR.  This can be used to check the built IR returns the same output as the model. |
| inference\_int8 | Integer inference with the compiled and quantized model.  The model must be compiled before executing this function. |
| inference\_int8\_input\_dict | Same as "inference\_int8", but get a dictionary input which has a form of {node name: node input} instead.  This can be used for models with multiple inputs. |
| cal\_ops | Return the number of add/multiplication operations in the build Mobilint IR.  This can be reduced in later optimization steps. |
| to | Set the operating device (CPU or GPU). |

### Method Details

Table 7‑3. Model\_Dict.\_\_init\_\_

| Parameter | Type | Description |
| --- | --- | --- |
| model | string or model class of the corresponding framework | Model path or model instance. The following cases are supported:  Using backend="onnx" and a ONNX model path,  Using backend="torchscript" and a PyTorch model path,  Using backend="tf1" or "tf2" and a fronzen TensorFlow PB graph. |
| backend | string (optional) | Which framework to use to get the Mobilint IR.  It must be one of "onnx", "tf1", "tf2", and "torchscript".  They correspond to deep learning frameworks as follows:  "onnx": ONNX,  "tf1" and "tf2": TensorFlow,  Defaults to "onnx". |
| input\_shape | tuple or list (optional) | Input shape in HWC. Required only for using PyTorch model. |
| device | string (optional) | Device to be used for compile and inerence. Either "cpu" or "gpu".  Defaults to "cpu". |

Table 7‑4. Model\_Dict.compile

| Parameter | Type | Description |
| --- | --- | --- |
| calib\_data\_path | string | A path to the calibration dataset. It can be either of a path to the text file (or json) containing the paths to the pre-processed numpy files or a directory containing the pre-processed numpy files. |
| save\_path | string | Filename of the resulting .mxq. |
| model\_nickname | string (optional) | Model nickname used in qubee. It is used in qubee to facilitate quicker recompilation of the same models.  Qubee stores prior optimization information under this nickname, enabling it to locate and utilize the previously compiled results for faster processing.  It is auto-generated from the model's base name, if not provided.  For instance, a model "/workspace/onnx/resnet50.onnx" results in "resnet50".  If not derivable, "temporary" is the default nickname. |
| quantize\_method | string (optional) | Quantization method to determine the scale parameter in the quantization.  Currently, "Max", "Percentile", "MaxPercentile" and "KL" are supported.  Defaults to "Percentile". |
| quantize\_percentile | float (optional) | Percentile used for the quantization method "Percentile" and "MaxPercentile".  This should be between 0 and 1. (Ex. 0.999, 0.9999).  Defaults to 0.9999. |
| topk\_ratio | float (optional) | It is used for quantization method "maxpercentile". Defaults to 0.  The larger this value is, the more data is used for calibration.  This should be between 0 and 1, but using a value of 0.01 or less is recommended. |
| smooth\_factor | float (optional) | Smoothing factor that is required for Gaussian kernel construction on KL divergence estimation.  Defaults to 1.6. |
| is\_quant\_ch | bool (optional) | Use multi-channel quantization if True. Defaults to False. |
| optimization | bool (optional) | If True, it compiles the model with optimization process.  If False, qubee uses  previous optimization information when stored in previous compiling.  (Nickname should be the same.) It must be set to True on the first compile.  Defaults to True. |
| optimization\_level | int (optional) | Optimization level in the compiler.  If optimization level is high, NPU inference  could be faster, but it takes more time for compiling. (Recommend: 3~6.)  Defaults to 5. |
| save\_sample | bool (optional) | If True, create the "sampleInOut" folder in the current directory and store the input and output binary files in it.  Defaults to False. |
| use\_random\_calib | bool (optional) | If True, it compiles the given model with random calibration data.  This is just used to check if the model is compilable without making a calibration data.  Defaults to False. |
| cpu\_offload | bool (optional) | Use CPU offloading for NPU inference if True. Defaults to False. |
| quant\_output | string (optional) | Quantization method that applied to the output layer. "layer", "ch" and "sigmoid" options are available.  If "layer", per-layer quantization is applied to the output layer. If is\_quant\_ch is true, then the computed quantization scale for each channel of the output layer will be merged into single value.  If "ch", per-channel quantization is applied to the output layer. This option is valid only when is\_quant\_ch is true.  If "sigmoid", assign quantization scale that computed with sigmoid function.  Defaults to "layer". |
| adaq\_useadaquant | bool (optional) | If True, enable the finetuning with AdaQuant after quantization.  Defaults to False. |
| adaq\_weightDeltaLR | float (optional) | Learning rate for finetuning weight delta(weight update) of AdaQuant. (Recommend: 1e-6 ~ 5e-5)  Defaults to 0. |
| adaq\_biasDeltaLR | float (optional) | Learning rate for finetuning bias delta(bias update) of AdaQuant. (Recommend: weightDeltaLR/10 ~ weightDeltaLR/2)  Defaults to 0. |
| adaq\_weightScaleLR | float (optional) | Learning rate for finetuning weight quantization scale of AdaQuant.  Defaults to 0. |
| adaq\_biasScaleLR | float (optional) | Learning rate for finetuning bias quantization scale of AdaQuant.  Defaults to 0. |
| adaq\_actScaleLR | float (optional) | Learning rate for finetuning activation quantization scale of AdaQuant.  Defaults to 0. |
| adaq\_batchSize | int (optional) | Batch size for running AdaQuant.  Defaults to 16. |
| adaq\_epoch | int (optional) | Epochs for repeating AdaQuant update.  Defaults to 10. |

Table 7‑5. Model\_Dict.inference

| Parameter | Type | Description |
| --- | --- | --- |
| input\_tensor | numpy.array  torch.Tensor  Dict[string, numpy.array or torch.Tensor]  List[numpy.array or torch.Tensor] | Input tensor with layout BCHW. |
| cast\_cpu | bool (optional) | If True, enable CPU casting on full precision inference.  Defaults to False. |

Table 7‑6. Model\_Dict.inference\_int8

| Parameter | Type | Description |
| --- | --- | --- |
| input\_tensor | torch.Tensor or np.ndarray | Input tensor with layout BCHW. |

Table 7‑7. Model\_Dict.inference\_int8\_input\_dict

| Parameter | Type | Description |
| --- | --- | --- |
| input\_dict | Dict[str, torch.Tensor or np.ndarray] | Dictionary that contains input information such as {input node name: input tensor}. |

Table 7‑8. Model\_Dict.to

| Parameter | Type | Description |
| --- | --- | --- |
| device | string | Target device to use, which must be one of "cpu", "gpu", "cuda". |

## Function: mxq\_compile

Compile a given model directly without creating an instance of "Model\_Dict".

Table 7‑9. mxq\_compile

| Parameter | Type | Description |
| --- | --- | --- |
| model | string or model instance | Model path or model instance. Model should be instance for the following cases:  Using backend="onnx" and a ONNX model path,  Using backend="torchscript" and a PyTorch model,  Using backend="tf1" or "tf2" and a fronzen TensorFlow PB graph. |
| calib\_data\_path | string | A path to the calibration dataset. It can be either of a path to the text file (or json) containing the paths to the pre-processed numpy files or a directory containing the pre-processed numpy files. |
| model\_nickname | string (optional) | Model nickname used in qubee. It is used in qubee to facilitate quicker recompilation of the same models.  Qubee stores prior optimization information under this nickname, enabling it to locate and utilize the previously compiled results for faster processing.  It is auto-generated from the model's base name, if not provided.  For instance, a model "/workspace/onnx/resnet50.onnx" results in "resnet50".  If not derivable, "temporary" is the default nickname. |
| save\_path | string (optional) | Filename of the resulting .mxq.  If it is None, then it is set to "model\_nickname".mxq  Defaults to None. |
| input\_shape | tuple or list (optional) | Input shape in HWC. Required only for using PyTorch model and backend="torchscript". |
| backend | string (optional) | Which framework to use to get the Mobilint IR.  It must be one of "onnx", "tf1", "tf2", and "torchscript".  They correspond to deep learning frameworks as follows:  "onnx": ONNX,  "tf1" or "tf2": TensorFlow, Keras  "torchscript": PyTorch  Defaults to "onnx". |
| device | string (optional) | Device to be used for compile and inerence. Either "cpu" or "gpu".  Defaults to "cpu". |
| quantize\_method | string (optional) | Quantization method to determine the scale parameter in the quantization.  Currently, "Max", "Percentile", "MaxPercentile" and "KL" are supported.  Defaults to "Percentile". |
| quantize\_percentile | float (optional) | Percentile used for the quantization method "Percentile" and "MaxPercentile".  This should be between 0 and 1. (Ex. 0.999, 0.9999)  Defaults to 0.99995. |
| topk\_ratio | float (optional) | It is used for quantization method "maxpercentile". Defaults to 0.  The larger this value is, the more data is used for calibration.  This should be between 0 and 1, but using a value of 0.01 or less is recommended. |
| smooth\_factor | float (optional) | Smooth factor for Gaussian kernel construction, which is required on KL divergence estimation.  Defaults to 1.6. |
| is\_quant\_ch | bool (optional) | Use multi-channel quantization if True. Defaults to False. |
| optimization | bool (optional) | If True, it compiles the model with optimization process. If false, qubee uses  previous optimization information when stored in previous compiling.  (Nickname should be the same.) It must be set to True on the first compile.  Defaults to True. |
| optimization\_level | int (optional) | Optimization level in the compiler. If optimization level is high, NPU inference  could be faster, but it takes more time for compiling. (Recommend: 3~6)  Defaults to 5. |
| save\_sample | bool (optional) | If True, create the "sampleInOut" folder in the current directory and store the input and output binary files in it.  Defaults to False. |
| use\_random\_calib | bool (optional) | If True, it compiles the given model with random calibration data.  This is just used to check if the model is compilable without making a calibration data.  Defaults to False. |
| cpu\_offload | bool (optional) | Use CPU offloading for NPU inference if True. Defaults to False. |
| quant\_output | string (optional) | Quantization method that applied to the output layer. "layer", "ch" and "sigmoid" options are available.  If "layer", per-layer quantization is applied to the output layer. If is\_quant\_ch is true, then the computed quantization scale for each channel of the output layer will be merged into single value.  If "ch", per-channel quantization is applied to the output layer. This option is valid only when is\_quant\_ch is true.  If "sigmoid", assign quantization scale that computed with sigmoid function.  Defaults to "layer". |
| adaq\_useadaquant | bool (optional) | If True, enable the finetuning with AdaQuant after quantization.  Defaults to False. |
| adaq\_weightDeltaLR | float (optional) | Learning rate for finetuning weight delta(weight update) of AdaQuant. (Recommend: 1e-6 ~ 5e-5)  Defaults to 0. |
| adaq\_biasDeltaLR | float (optional) | Learning rate for finetuning bias delta(bias update) of AdaQuant. (Recommend: weightDeltaLR/10 ~ weightDeltaLR/2)  Defaults to 0. |
| adaq\_weightScaleLR | float (optional) | Learning rate for finetuning weight quantization scale of AdaQuant.  Defaults to 0. |
| adaq\_biasScaleLR | float (optional) | Learning rate for finetuning bias quantization scale of AdaQuant.  Defaults to 0. |
| adaq\_actScaleLR | float (optional) | Learning rate for finetuning activation quantization scale of AdaQuant.  Defaults to 0. |
| adaq\_batchSize | int (optional) | Batch size for running AdaQuant.  Defaults to 16. |
| adaq\_epoch | int (optional) | Epochs for repeating AdaQuant update.  Defaults to 10. |

### Tips for choosing quantization methods

"Percentile" and "MaxPercentile" quantization methods each take a hyperparameter called *percentile*. An increase in this value corresponds to a broader quantization interval. To elaborate further, a higher *percentile* results in reduced overflow, albeit at the expense of accuracy.

The "MaxPercentile" method determines the percentile value from data that has been filtered once. As a result, a lower *percentile* is needed for "MaxPercentile" compared to the "Percentile" method. For instance, for the "Percentile" method, we suggest using a value of 0.9999 to 0.999999. For the "MaxPercentile" method, we recommend *percentile* between 0.9 and 0.9999.

## Function: make\_calib

From given images and preprocessing configuration, create the preprocessed numpy files and a txt file containing their paths.

Table 7‑10. make\_calib

| Parameter | Type | Description |
| --- | --- | --- |
| args\_pre | string or Dict | Path to a Yaml file or dictionary containing preprocessing configuration information.  Refer to 7.4. for details. |
| data\_dir | string | Directory of data to be used for calibration. |
| save\_dir | string | Directory to save the pre-processed numpy files and txt file which contains their paths. |
| save\_name | string (optional) | Name for resulting files.  Numpy files will be saved under {save\_dir}/{save\_name}\_npy directory.  Text file will be saved in {save\_dir}/{save\_name}.txt.  If it is not provided, it is set to the basename of data\_dir. |
| anno\_json | string (optional) | Path to an annotation json file for COCO format.  When provided, make\_calib function randomly selects samples considering class balance.  Defaults to None. |
| file\_format | string (optional) | Filename format using image\_idx.  Defaults to '%012d.jpg'. |
| max\_size | int (optional) | Maximum size of the resulting calibration data.  Defaults to -1, which means no limit on the number of the calibration data. |
| remove\_npy | bool (optional) | If True, remove pre-existing numpy files.  Defaults to False. |
| seed | int (optional) | Random seed.  Defaults to 2023. |
| save\_calib\_msg | bool (optional) | If True, save calibration data dictionary as MSGpack file.  Defaults to False. |
| msg\_path | string (optional) | Path to save MSGpack file  If not provided, it automatically generate the path with dataname and number of calibration data.  Defaults to None. |

## Fuction: make\_calib\_man

From given images and manually written function that takes an image path as input, create the preprocessed numpy files and a txt file containing their paths.

Table 7‑11. make\_calib\_man

| Parameter | Type | Description |
| --- | --- | --- |
| pre\_ftn | Callable | Pre-processing function that takes an image path as input. |
| data\_dir | string | Directory of data to be used for calibration. |
| save\_dir | string | Directory to save the pre-processed numpy files and txt file which contains their paths. |
| save\_name | string (optional) | Name for resulting files.  Numpy files will be saved under {save\_dir}/{save\_name}\_npy directory.  Text file will be saved in {save\_dir}/{save\_name}.txt.  If it is not provided, it is set to the basename of data\_dir. |
| anno\_json | string (optional) | Path to an annotation json file for COCO format.  When provided, make\_calib function randomly selects samples considering class balance.  Defaults to None. |
| file\_format | string (optional) | Filename format using image\_idx.  Defaults to '%012d.jpg'. |
| max\_size | int (optional) | Maximum size of the resulting calibration data.  Defaults to -1, which means no limit on the number of the calibration data. |
| remove\_npy | bool (optional) | If True, remove pre-existing numpy files.  Defaults to False. |
| seed | int (optional) | Random seed.  Defaults to 2023. |
| save\_calib\_msg | bool (optional) | If True, save calibration data dictionary as MSGpack file.  Defaults to False. |
| msg\_path | string (optional) | Path to save MSGpack file  If not provided, it automatically generate the path with dataname and number of calibration data.  Defaults to None. |

Example codes for using these functions are provided in the [##Preparing Calibration Data](#Preparing Calibration Data) section.

## Pre-processing Configurations

qubee supports the following pre-processing functions to make calibration data.

Table 7‑12. Pre-processing function API

| Pre-processing Type | Description |
| --- | --- |
| GetImage | Get image tensor from image path using cv2 backend or image tensor.  Note that this should be at the top of the list. |
| Pad | Pad image tensor. |
| Normalize | Normalize image tensor. |
| ResizeTorch | Resize the input image to the given size using [torchvision.transforms.functional.resize](https://pytorch.org/vision/0.14/generated/torchvision.transforms.functional.resize.html?highlight=resize#torchvision.transforms.functional.resize) |
| Resize | Resize image tensor to the given size using [cv2.resize](https://docs.opencv.org/4.6.0/da/d54/group__imgproc__transform.html#ga47a974309e9102f5f08231edc7e7529d). |
| CenterCrop | Center crop the image tensor. |
| SetOrder | Set the order of axes of the given image tensor.  Note that this should be at the very end. |

You can write a yaml file as follows:

|  |
| --- |
| [Pre-processing Type]  [Parameter]: [Argument]  ... |

|  |
| --- |
| *# Example*  GetImage:  to\_float32: **false**  channel\_order: RGB  ResizeTorch:  size: [256, 256]  interpolation: blinear  CenterCrop:  size: [224, 224]  Normalize:  mean: [0.485, 0.456, 0.406]  std: [0.229, 0.224, 0.225]  to\_float: **true**  SetOrder:  shape: HWC |

### Pre-processing Parameters

Table 7‑13. GetImage

| Parameter | Type | Description |
| --- | --- | --- |
| to\_float32 | bool (optional) | If True, set dtype as float32.  Defaults to False. |
| channel\_order | string (optional) | Channel order to load. Upper cases will be converted into lower cases. Defaults to "bgr". |

Table 7‑14. Pad

| Parameter | Type | Description |
| --- | --- | --- |
| shape | Tuple[int] (optional) | Expected padding shape (h, w). Defaults to None. |
| size\_divisor | int (optional) | Pad images so that the the resulting image's width and height are divisible by size\_divisor.  Defaults to None. |
| pad\_val | float (optional) | Values to be filled in padding areas when padding\_mode is 'constant'.  Defaults to 0. |
| right\_bottom | bool (optional) | If True, it only bads to right and bottom. Defaults to False. |

Table 7‑15. Normalize

| Parameter | Type | Description |
| --- | --- | --- |
| mean | List[float] or np.ndarray | Normalization mean. |
| std | List[float] or np.ndarray | Normalization standard deviation. |
| to\_float | bool (optional) | Normalize image between [0, 255] into [0, 1] by dividing by 255 before normalizing with the mean and std.  Defaults to False. |

Table 7‑16. ResizeTorch

| Parameter | Type | Description |
| --- | --- | --- |
| size | List[int] | Desired output size, i.e., height and width. |
| interpolation | string | Interpolation method,  accepted values are "nearest", "bilinear", "bicubic", "box", "hamming", "lanczos". |

Table 7‑17. Resize

| Parameter | Type | Description |
| --- | --- | --- |
| img\_scale | float or Tuple[int, int] | The scaling factor or maximum size (h, w).  If it is a float number, then the image will be rescaled by this factor, else if it is a tuple of 2 integers, then the image will be rescaled as large as possible within the scale. |
| keep\_ratio | bool | Whether to keep the aspect ratio when resizing the image. Defaults to False. |
| interpolation | string | Interpolation method,  accepted values are "nearest", "bilinear", "bicubic", "area", "lanczos". |

Table 7‑18. CenterCrop

| Parameter | Type | Description |
| --- | --- | --- |
| size | List[int] | Desired output height and width. |

Table 7‑19. SetOrder

| Parameter | Type | Description |
| --- | --- | --- |
| shape | string | Desired data layout format, accepted values are "HWC", "CHW", "BHWC", "BCHW".  Defaults to "HWC". |

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