

Smart Factory – Model Training

Objectives

- Al model을 training하는 방법을 학습한다
 - 직접 수집한 불량/양품 데이터로 AI 모델을 트레이닝하고 application에 적용할 수 있다
 - 원하는 기능을 수행하는 AI model을 training하고 프로젝트에 적용할 수 있다
 - OTX(OpenVINIO Training Extension) 사용 방법을 알 수 있다
 - Hyperparameter에 따른 성능 변화를 관찰 한다

Software Dependencies

- Ubuntu 22.04
- Python 3.10
 - sudo apt install python3 python3-dev python3-venv
- OpenVINO Training Extensions
- Refer <u>otx.md</u>

```
sudo apt-get install g++ freeglut3-dev build-essential libx11-dev
libxmu-dev libxi-dev libglu1-mesa libglu1-mesa-dev gcc-multilib dkms
mesa-utils
```

OpenVINO Training Extensions(OTX)

- Low-code transfer learning framework for Computer Vision
- Diverse combinations of model architectures, learning methods, and task types based on PyTorch and OpenVINO toolkit
- Provides model templates
- Supports Hyper Parameter Optimization(HPO)



Nvidia GPU

Python 3.8

Cuda 11.7 에서만 하기 내용으로 동작함.

Disable Nouveau & Install nVidia GPU driver

```
$ lsmod|grep nouveau
                  2285568 20
nouveau
        16384 1 nouveau
mxm wmi
drm ttm helper 16384 1 nouveau
            86016 2 drm ttm helper, nouveau
ttm
drm kms helper 307200 1 nouveau
i2c_algo_bit 16384 1 nouveau
drm
                   618496 11
drm kms helper,drm ttm helper,ttm,nouveau
video
                   61440 1 nouveau
                   32768 3 wmi bmof, mxm wmi, nouveau
wmi
```

```
sudo su
echo -e "\n\nblacklist nouveau" >> /etc/modprobe.d/blacklist.conf

echo "options nouveau modeset=0" > /etc/modprobe.d/nouveau-kms.conf
update-initramfs -u
reboot

sudo ubuntu-drivers autoinstall
```

Verify nVidia GPU driver installation

```
$ glxinfo|grep -i "opengl renderer"
OpenGL renderer string: NVIDIA GeForce GTX 1660/PCIe/SSE2
```

```
$ nvidia-smi
Sat Aug 5 12:07:21 2023
                                 Driver Version: 535.86.05 CUDA Version: 12.2
 NVIDIA-SMI 535.86.05
                      Persistence-M | Bus-Id
                                                      Disp.A | Volatile Uncorr. ECC
                                                Memory-Usage | GPU-Util Compute M.
             Perf
                          Pwr:Usage/Cap |
  Fan Temp
   0 NVIDIA GeForce GTX 1660
                                         00000000:01:00.0 On |
                                                                               N/A
       54C
                            50W / 130W |
                                           3403MiB / 6144MiB |
                                                                  98%
                                                                           Default
                                                                               N/A
  Processes:
  GPU
        GI
             CI
                      PID
                                                                        GPU Memory
                                  Process name
        ID
             ID
                              G /usr/lib/xorg/Xorg
        N/A N/A
                     1191
                                                                            428MiB |
        N/A N/A
                     1524
                              G /usr/bin/gnome-shell
                                                                             88MiB |
        N/A N/A
                                  ...irefox/2952/usr/lib/firefox/firefox
                                                                            173MiB |
        N/A N/A
                     5327
                                  ...sion, SpareRendererForSitePerProcess
                                                                            103MiB
                                  ...otx-classification/.otx/bin/python3
```

Install CUDA 11.7

wget

https://developer.download.nvidia.com/compute/cuda/11.7.0/1
ocal installers/cuda 11.7.0 515.43.04 linux.run

sudo sh cuda_11.7.0_515.43.04_linux.run



Existing package manager installation of the driver found. It is strongly recommended that you remove this before continuing.

Abort

Continue





Install CUDA 11.7 (Cont'd)

```
Edit .bashrc
export PATH=/usr/local/cuda-11.7/bin:$PATH
export LD_LIBRARY_PATH=/usr/local/cuda-11.7/lib64:$LD_LIBRARY_PATH

Or
echo "export PATH=/usr/local/cuda-11.7/bin:\$PATH" >> ~/.bashrc
echo "export LD_LIBRARY_PATH=/usr/local/cuda-11.7/lib64:\$LD_LIBRARY_PATH" >> ~/.bashrc
sudo reboot
```

OTX Install

```
# Create a virtual env.
python -m venv .otx

# Activate virtual env.
source .otx/bin/activate
```

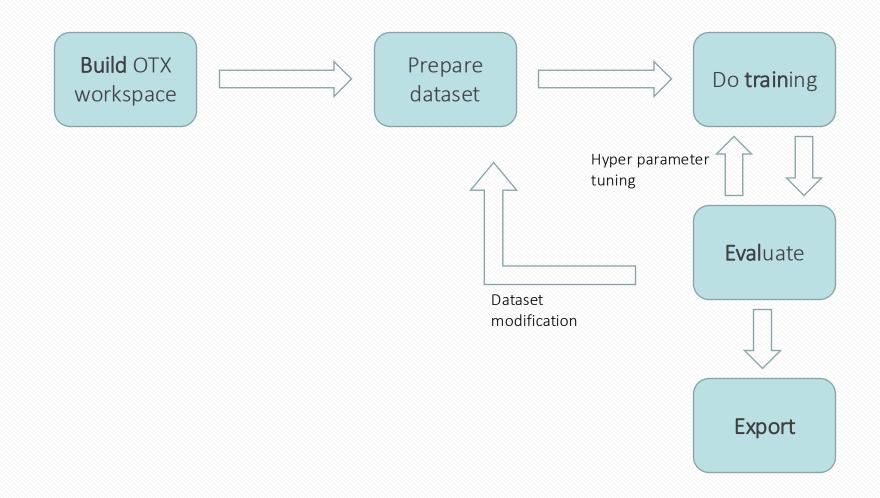
```
pip install wheel setuptools

# install command for torch==1.13.1 for CUDA 11.7:

pip install torch==1.13.1 torchvision==0.14.1 --extra-index-url <a href="https://download.pytorch.org/whl/cu117">https://download.pytorch.org/whl/cu117</a>

pip install otx[full]
```

OTX Training Steps



OTX find

• 'find' gives supported templates list

```
otx find
otx find | awk '{print $2}' | uniq
otx find --task classification
```

```
ACTION_CLASSIFICATION
ACTION_DETECTION
CLASSIFICATION
ANOMALY_CLASSIFICATION
ANOMALY_DETECTION
ANOMALY_SEGMENTATION
ROTATED_DETECTION
DETECTION
INSTANCE_SEGMENTATION
SEGMENTATION
```

OTX Classification Example

• Download the **flower_photos** dataset

```
find ./ -maxdepth 2 -type d | while read -r dir; do printf "%s:\t" "$dir"; find "$dir" -type f | wc
-1; done

./flower_photos/: 3671
./flower_photos/dandelion: 898
./flower_photos/tulips: 799
./flower_photos/daisy: 633
./flower_photos/sunflowers: 699
./flower_photos/roses: 641
```

OTX Classification Example - build

Make sure otx virtual environment was loaded

```
source .otx/bin/activate
```

Create a workspace for classification

```
(.otx)$ mkdir ~/workspace/otx-flowers && cd $_
(.otx)$ otx build --train-data-roots ./flower_photos/ --model MobileNet-V3-large-1x --workspace
./classification-task
```

Explore the workspace

```
(.otx)$ cd classification-task
(.otx)$ cat template.yaml
(.otx)$ ds_count ./splitted_dataset 1
./splitted_dataset/: 3670
./splitted_dataset/train: 2936
./splitted_dataset/val: 734
```

Manage Your Favorite Functions

You can create a bash function for your own comfort

```
mkdir -p ~/bin/
echo -e "\n\nsource ~/bin/my_funcs.sh" >> ~/.bashrc
touch ~/bin/my_funcs.sh
```

```
cat ~/bin/my_funcs.sh

ds_count() {
    find $1 -maxdepth $2 -type d | \
      while read dir
         do print "%s:\t" "$dir"
         find "$dir" -type f | wc -l
      done
}
```

```
# Log out/in or source the bash resource file
source ~/.basrhrc
# Newly defined dataset counting function should work
ds_count ./ 2
```

Troubleshoot - NotADirectoryError

```
$ otx build --train-data-roots ~/dataset/flower photos --model EfficientNet-V2-S --workspace classification-task
[*] Detected dataset format: imagenet
Traceback (most recent call last):
 File "/home/intel/workspace/otx-classification/.otx/lib/python3.10/site-packages/datumaro/components/dataset.py", line 728, in import from
  env.make extractor(src conf.format, src conf.url, **extractor kwargs)
 File "/home/intel/workspace/otx-classification/.otx/lib/python3.10/site-packages/datumaro/components/environment.py", line 283, in make extractor
  return self.extractors.get(name)(*args, **kwargs)
 File "/home/intel/workspace/otx-classification/.otx/lib/python3.10/site-packages/datumaro/plugins/data formats/imagenet.py", line 40, in init
  self. categories = self. load categories(path)
 File "/home/intel/workspace/otx-classification/.otx/lib/python3.10/site-packages/datumaro/plugins/data_formats/imagenet.py", line 47, in _load_categories
  raise NotADirectoryError(errno.ENOTDIR, os.strerror(errno.ENOTDIR), path)
NotADirectoryError: [Errno 20] Not a directory: '/home/intel/dataset/flower photos'
The above exception was the direct cause of the following exception:
Traceback (most recent call last):
```

Troubleshoot – **NotADirectoryError**(Cont'd)

Trigger PDB Python debugger

```
import pdb; pdb.set_trace()
```

```
def _load_categories(self, path):

label_cat = LabelCategories()

for dirname in sorted(os.listdir(path)):

if not os.path.isdir(os.path.join(path, dirname)):

import pdb; pdb.set_trace()

ratse NotAbtrectoryError(error.ENOIDIR, os.strerror(error));

if dispare | TracecatPath TMACE_DID_NO_LABEL:
```

```
    Updated: classificatoin-task/compression_config.json

   Detected dataset format: imagenet
  /home/intel/workspace/otx-classification/.otx/lib/python3.10/site-packages/datumaro/plugins/data
> raise NotADirectoryError(errno.ENOTDIR, os.strerror(errno.ENOTDIR), path)
(Pdb) l
           def load categories(self, path):
43
               label cat = LabelCategories()
               for dirname in sorted(os.listdir(path)):
                   if not os.path.isdir(os.path.join(path, dirname)):
47
                        import pdb; pdb.set trace()
                        raise NotADirectoryError(errno.ENOTDIR, os.strerror(errno.ENOTDIR), path)
                   if dirname != ImagenetPath.IMAGE DIR NO LABEL:
                        label cat.add(dirname)
51
               return {AnnotationType.label: label cat}
52
           def load_items(self, path):
(Pdb) p path
'/home/intel/dataset/flower photos'
(Pdb) p dirname
'LICENSE.txt'
(Pdb) os.path.isdir(os.path.join(path, dirname))
False
(Pdb) p os.path.join(path, dirname)
 /home/intel/dataset/flower_photos/LICENSE.txt'
```

OTX Classification Example - train

Start training

```
(.otx)$ otx train
2023-08-01 11:37:14,947 | INFO : Epoch [21][46/46] | lr: 5.122e-03, eta: 0:29:50, time:
0.254, data time: 0.109, memory: 1855, current iters: 965, loss: 0.0109, sharpness: 0.0383,
max loss: 0.0492
[>>>>>>>>>>> task/s, elapsed: 2s, ETA:
0 s
2023-08-01 11:37:16,826 | INFO :
Early Stopping at :20 with best accuracy: 0.9618528747558593
2023-08-01 11:37:16,826 | INFO : Exp name: outputs/20230801 113219 train/logs
2023-08-01 11:37:16,826 | INFO : Epoch(val) [21][46] accuracy top-1: 0.9578, accuracy top-5:
1.0000, daisy accuracy: 0.9606, dandelion accuracy: 0.9944, roses accuracy: 0.9457, sunflowers
accuracy: 0.9640, tulips accuracy: 0.9187, mean accuracy: 0.9567, accuracy: 0.9578, current iters:
966
2023-08-01 11:37:16,828 | INFO : MemCacheHandlerBase uses 0 / 0 (0.0%) memory pool and store 0
items.
2023-08-01 11:37:18,115 | INFO : called save model
2023-08-01 11:37:18,283 | INFO: Final model performance: Performance(score: 0.9618528747558593,
dashboard: (18 metric groups))
2023-08-01 11:37:18,283 | INFO : train done.
otx train time elapsed: 0:04:59.112100
otx train CLI report has been generated: outputs/20230801 113219 train/cli report.log
```

Troubleshoot – loss: nan



Troubleshoot – **loss: nan** (Cont'd)

- Try to adjust learning rate if nan happens during a training
 - calculation result diverged
 - divided by zero

$$w' = w - \alpha \frac{\partial J}{\partial w}$$
 $b' = b - \alpha \frac{\partial J}{\partial b}$ learning rate

$$b' = b - \alpha \frac{\partial J}{\partial b}$$
learning rate

[template.yaml]

learning rate:

default value: 0.0058

auto hpo_state: POSSIBLE

```
# Maximum learning rate: 1.0 -- configuration.yaml
$ otx train params --learning_parameters.learning_rate 1.0 --learning_parameters.num_iters 1
# Minimum learning rate: 1e-07 – configuration.yaml
$ otx train params --learning_parameters.learning_rate 1e-07 --learning_parameters.num_iters 1
```

Troubleshoot – **loss: nan** (Cont'd)

- Try to adjust batch size
 - depends on GPU's hardware capability
 - smaller batch size will take longer time to achieve high accuracy

[template.yaml]

batch_size:

default_value: 64

auto_hpo_state: POSSIBLE

```
# Half size of default batch size(64)
$ otx train params --learning_parameters.batch_size 32 --learning_parameters.num_iters 1

# Try more half
$ otx train params --learning_parameters.batch_size 16 --learning_parameters.num_iters 1
```

OTX Classification Example - train

Training outputs

OTX Classification Example – eval(optional)

• Evaluate trained model for test dataset

```
(.otx)$ otx eval --test-data-roots ./splitted_dataset/val --load-weights
./outputs/latest_trained_model/logs/best_epoch_13.pth
```

OTX Classification Example – export

Export trained model to OpenVINO format

```
# Sotx export

...
Find more information about API v2.0 and IR v11 at
https://docs.openvino.ai/latest/openvino_2_0_transition_guide.html
[ SUCCESS ] Generated IR version 11 model.
[ SUCCESS ] XML file: /home/litcoder/workspace/otx-flowers/classification-
task/outputs/20230801_140718_export/logs/model.xml
[ SUCCESS ] BIN file: /home/litcoder/workspace/otx-flowers/classification-
task/outputs/20230801_140718_export/logs/model.bin

2023-08-01 14:07:26,630 - mmdeploy - INFO - Successfully exported OpenVINO model:
outputs/20230801_140718_export/logs/model_ready.xml
2023-08-01 14:07:26,749 | INFO : Exporting completed
```

OTX Classification Example – Application

• Test the trained model with any downloaded flower images

```
(.otx) $ wget
https://raw.githubusercontent.com/openvinotoolkit/openvino/master/samples/python/hello classificati
on/hello classification.py
(.otx) $ python hello classification.py ./outputs/20230801 140718 export/openvino/openvino.xml
./test/1.jpg "CPU"
[ INFO ] Creating OpenVINO Runtime Core
[ INFO ] Reading the model: ./outputs/20230801 140718 export/openvino/openvino.xml
[ INFO ] Loading the model to the plugin
[ INFO ] Starting inference in synchronous mode
[ INFO ] Image path: ./test/1.jpg
[ INFO ] Top 10 results:
[ INFO ] class id probability
[ INFO ] -----
                                   'daisy', 'dandelion', 'roses',
[ INFO ] 3 3.5824265
[ INFO ] 0 1.9845574
                                       'sunflowers', 'tulips'
[ INFO ] 2 -0.5692880
[ INFO ] 1 -2.5086317
                -2.9615641
[ INFO ] 4
[ INFO ]
[ INFO ] This sample is an API example, for any performance measurements please use the dedicated
benchmark app tool
```



AI TRAINING SERIES

Install Intel arc graphic deriver in ubuntu (1)

url: https://www.intel.co.kr/content/www/kr/ko/download/747008/intel-arc-graphics-driver-ubuntu.html



소개

이 릴리스는 인텔® Arc™ A-시리즈 그래픽의 Ubuntu* 23.04 및 22.04용 패키지 업데이트를 제공합니다. 세부 설명

인텔® Arc™ A-시리즈 그래픽용 Linux* 기반 운영 체제 배포용 범용 GPU(GPGPU) 기능을 활성화하기 위 해 인텔 소프트웨어 설치, 배포 및 업데이트하는 방법에 대한 설명서 및 지침은 여기 https://dqpudocs.intel.com/driver/client/overview.html 찾을 수 있습니다.

OS 지원

- 우분투* 22.04
- 우분투* 23.04

Install Intel arc graphic deriver in ubuntu (2)

url: https://dgpu-docs.intel.com/driver/client/overview.html

Ubuntu 버전 24.10

Installing Client GPUs on Ubuntu Desktop 24.10

Support for Lunar Lake and initial support for Battle Mage has been backported from kernel version 6.12 to version 6.11, which is included in Ubuntu 24.10. However, as this version of Ubuntu does not include the latest compute and media-related packages, we offer the intel-graphics Personal Package Archive (PPA). The PPA provides early access to newer packages, along with additional tools and features such as EU debugging.

Use the following commands to install the intel-graphics PPA and the necessary compute and media packages:

```
apt-get update
apt-get install -y software-properties-common

# Add the intel-graphics PPA for 24.10
add-apt-repository -y ppa:kobuk-team/intel-graphics

# Install the compute-related packages
apt-get install -y libze-intel-gpu1 libze1 intel-ocloc intel-opencl-icd clinfo

# Install the media-related packages
apt-get install -y intel-media-va-driver-non-free libmfx1 libmfx-gen1.2 libvpl2 libvpl-tools libva-glx2 va-driver-all vainfc
```

Install Intel arc graphic deriver in ubuntu (3)

url: https://dgpu-docs.intel.com/driver/client/overview.html

Ubuntu 버전 24.04

Installing Client GPUs on Ubuntu Desktop 24.04 LTS

The Ubuntu 24.04 repositories contain compute packages for various Intel graphics products. To install those packages, use the following commands:

```
# Install the Intel graphics GPG public key
wget -q0 - https://repositories.intel.com/gpu/intel-graphics.key | \
    sudo gpg --yes --dearmor --output /usr/share/keyrings/intel-graphics.gpg

# Configure the repositories.intel.com package repository
echo "deb [arch=amd64,i386 signed-by=/usr/share/keyrings/intel-graphics.gpg] https://repositories.intel.com/gpu/ubuntu noble
    sudo tee /etc/apt/sources.list.d/intel-gpu-noble.list

# Update the package repository meta-data
sudo apt update

# Install the compute-related packages
apt-get install -y libzel intel-level-zero-gpu intel-opencl-icd clinfo
```

Install Intel arc graphic deriver in ubuntu (4)

url: https://dgpu-docs.intel.com/driver/client/overview.html

Ubuntu 버전 22.04

Installing Client GPUs on Ubuntu Desktop 22.04 LTS

The Ubuntu 22.04 repositories do not contain compute packages for various Intel graphics products. To install these packages, you can use Intel's dedicated package repository.

```
# Install the Intel graphics GPG public key

wget -q0 - https://repositories.intel.com/gpu/intel-graphics.key | \
sudo gpg --yes --dearmor --output /usr/share/keyrings/intel-graphics.gpg

# Configure the repositories.intel.com package repository

echo "deb [arch=amd64,i386 signed-by=/usr/share/keyrings/intel-graphics.gpg] https://repositories.intel.com/gpu/ubuntu jammy
sudo tee /etc/apt/sources.list.d/intel-gpu-jammy.list

# Update the package repository meta-data
sudo apt update

# Install the compute-related packages
apt-get install -y libzel intel-level-zero-gpu intel-opencl-icd clinfo
```

Install openvino training extension (otx)(1)

github: https://github.com/openvinotoolkit/training extensions

installation guide:

https://openvinotoolkit.github.io/training extensions/latest/guide/get started/installation.html

Otx 설치 가이드 페이지 이동

Installation

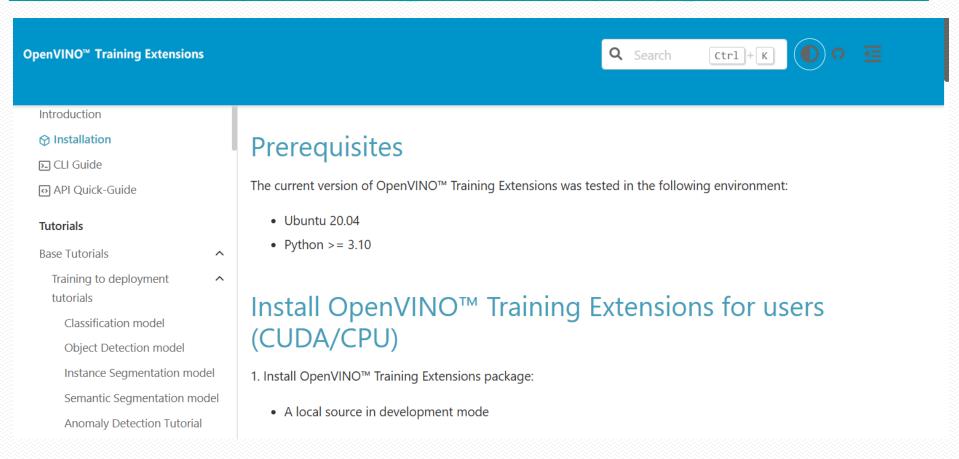
Please refer to the <u>installation guide</u>. If you want to make changes to the library, then a local installation is recommended.

- ► Install from PyPI
- ▶ Install from source

Install openvino training extension (otx)(2)

installation guide:

https://openvinotoolkit.github.io/training extensions/latest/guide/get started/installation.html



Install openvino training extension (otx)(3)

installation guide:

https://openvinotoolkit.github.io/training extensions/latest/guide/get started/installation.html

중요! - Intel arc gpu의 경우 반드시 XPU devices로 설치!!

Install OpenVINO™ Training Extensions for users (XPU devices)

1. Install OpenVINO™ Training Extensions from source to use XPU functionality.

Minimum requirements

Clone the training_extensions repository with the following command:
git clone https://github.com/openvinotoolkit/training_extensions.git
cd training_extensions

pip install -e '.[base]' --extra-index-url https://download.pytorch.org/whl/test/xpu

Note

Please, refer to the PyTorch official documentation guide to install prerequisites and resolve possible issues.

Install openvino training extension (otx)(4)

installation guide:

https://openvinotoolkit.github.io/training extensions/latest/guide/get started/installation.html

Install pytorch for Intel arc graphic

Install OpenVINO™ Training Extensions for users (XPU devices)

1. Install OpenVINO™ Training Extensions from source to use XPU functionality.

Minimum requirements

```
# Clone the training_extensions repository with the following command:
git clone https://github.com/openvinotoolkit/training_extensions.git
cd training_extensions
pip install -e '.[base]' --extra-index-url https://download.pytorch.org/whl/test/xpu
```

Note

Please, refer to the PyTorch official documentation guide to install prerequisites and resolve possible issues.

Install openvino training extension (otx)(5)

Getting Started on Intel GPU on pytorch:

https://pytorch.org/docs/stable/notes/get_start_xpu.html

Install pytorch for Intel arc graphic

Binaries

Platform Linux

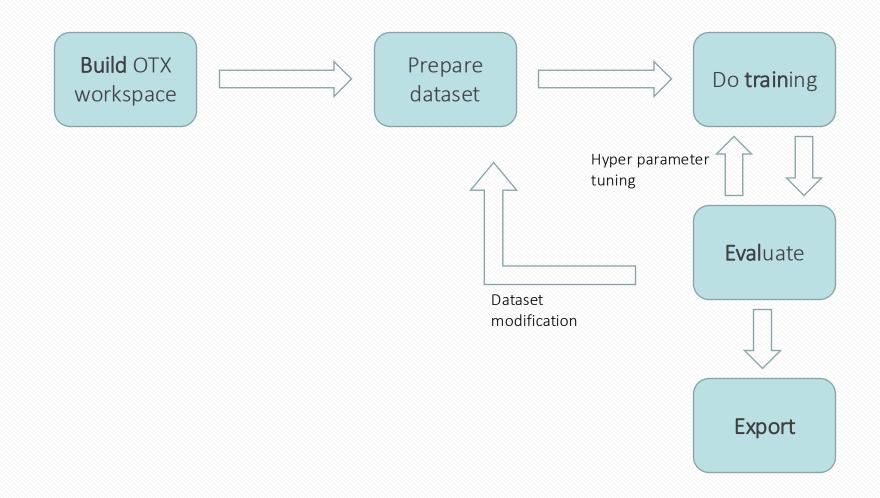
Now we have all the required packages installed and environment activated. Use the following commands to install pytorch, torchvision, torchaudio on Linux.

For preview wheels

pip3 install torch torchvision torchaudio --index-url https://download.pytorch.org/whl/test/xpu



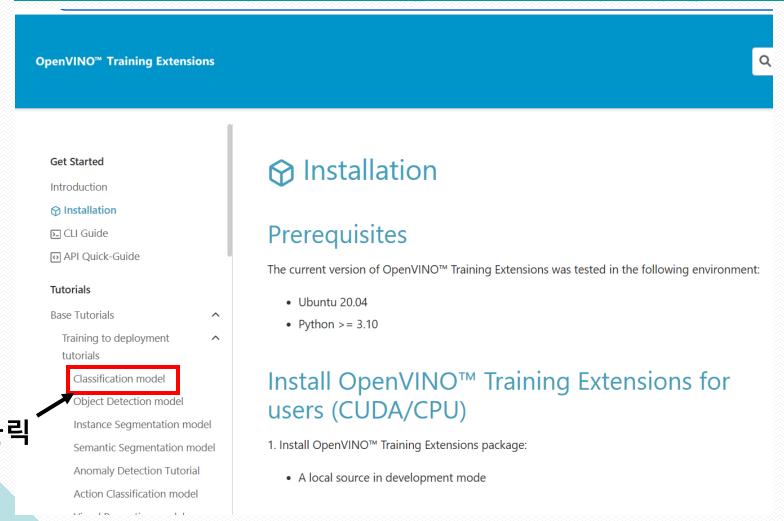
OTX Training Steps



Classification using OTX (1)

installation guide:

https://openvinotoolkit.github.io/training extensions/latest/guide/get started/installation.html



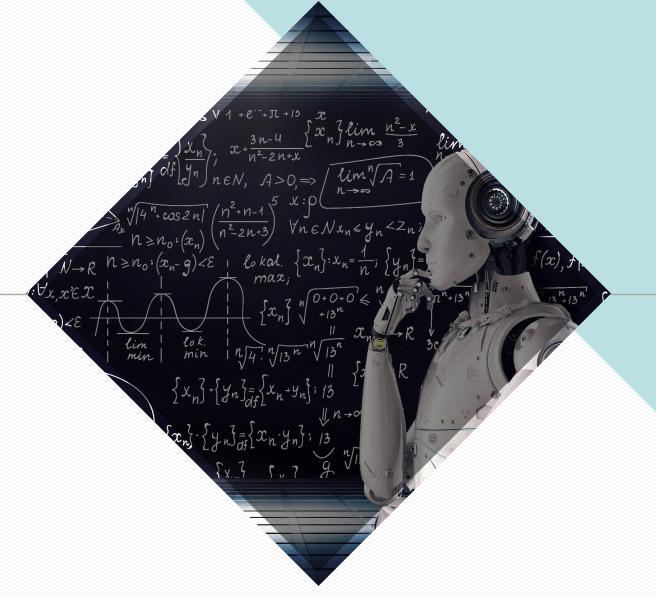
Homework #3 – Train your own AI models

- Collect dataset for Smart Factory defect classification
- Train your own model
 - Try all classification models
 - Tune hyper parameters for highest accuracy
 - Measure FPS(Frame Per Seconds: inference speed)
 - Use hw03 template.md for reporting out
- Modify inference code as necessary

Homework03

Classification model	Accuracy	FPS	Training time	Batch size	Learning rate	Other prams
EfficientNet-V2-S						
EfficientNet-B0						
DeiT-Tiny						
MobileNet-V3-large-1x						

FPS 측정 방법



THANK YOU