YOLOv9 Inferencing

하성욱 교수

```
✓ [1] import os, glob
      from IPython.display import Image
      from google.colab import drive, userdata
      HOME = os.getcwd()
      YOLO = os.path.join(HOME, 'yolov9')
      print(HOME)
      print(YOLO)
      /content
      /content/volov9
     # 구글 드라이브 마운트
      drive.mount('/content/drive')
     Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).
     pip install -q "openvino>=2023.3.0" "nncf>=2.8.1" "opencv-python" "seaborn" "pandas" "scikit-learn" "torch" "torchvision" "tqdm" --extra-index-url https://download.pytorch.org/whl/cpu
       Preparing metadata (setup.py) ... done
                                                 ----- 68.4/68.4 kB 3.4 MB/s eta 0:00:00
                                                     ----- 207.3/207.3 kB 10.1 MB/s eta 0:00:00
       Preparing metadata (setup.py) ... done
                         ----- 307.2/307.2 kB 29.4 MB/s eta 0:00:00
                                        ----- 4.2/4.2 MB 102.3 MB/s eta 0:00:00
                                             ----- 249.1/249.1 kB 25.3 MB/s eta 0:00:00
                                          ----- 76.0/76.0 kB 9.0 MB/s eta 0:00:00
                                                         ---- 119.4/119.4 kB 12.5 MB/s eta 0:00:00
       Building wheel for istyleson (setup.py) ... done
       Building wheel for grapheme (setup.pv) ... done
```

```
| Signature | Sign
```

PyTorch 모델을 OpenVINO IR로 변환

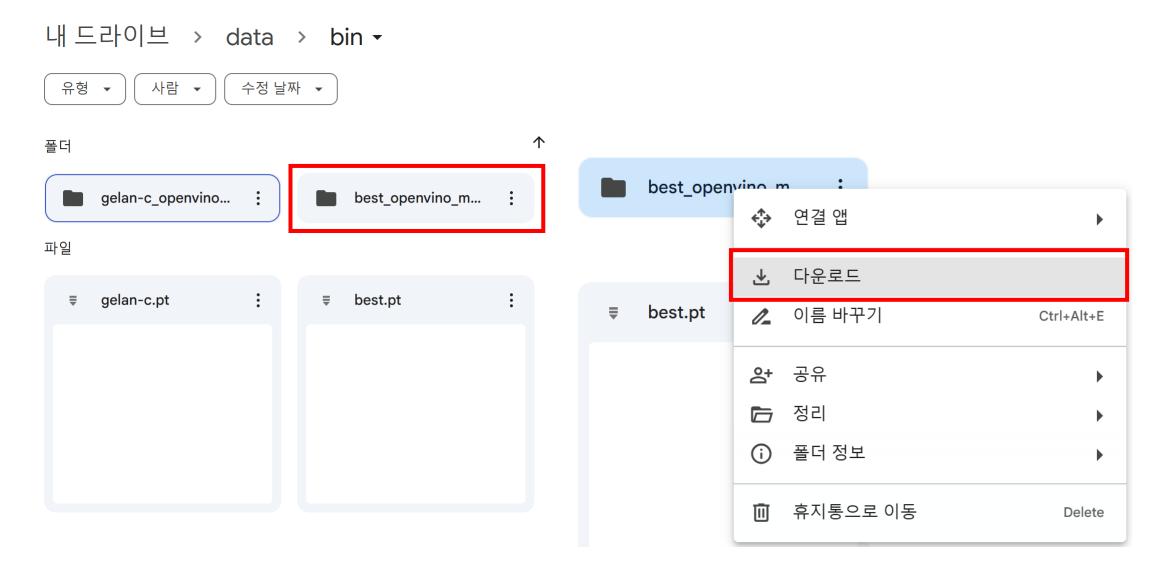
OpenVINO는 모델 변환 API를 제공한다. ov.convert_model 함수는 모델 객체와 모델을 분석하기 위한 입력을 받아서, ov.Model 인스턴스를 리턴한다. 리턴된 모델은 특정 장치용으로 로딩하거나 ov.save_model을 사용하여 다음 추론을 위해 저장될 수 있다.

```
[6] from models.experimental import attempt_load
     import torch
     import openvino as ov
    from models.yolo import Detect, DualDDetect
    from utils.general import yaml save, yaml load
    from pathlib import Path
     MODEL_DIR = Path("/content/drive/MyDrive/data/bin/")
    weights = MODEL_DIR / "best.pt"
    ov_model_path = MODEL_DIR / weights.name.replace(".pt", "_openvino_model") / weights.name.replace(".pt", ".xml")
    if not ov_model_path.exists():
         model = attempt_load(weights, device="cpu", inplace=True, fuse=True)
         metadata = {"stride": int(max(model.stride)), "names": model.names}
         model.eval()
        for k, m in model.named modules():
             if isinstance(m, (Detect, DualDDetect)):
                 m.inplace = False
                 m.dynamic = True
                 m.export = True
         example_input = torch.zeros((1, 3, 640, 640))
         model(example input)
        ov model = ov.convert model(model, example input=example input)
```

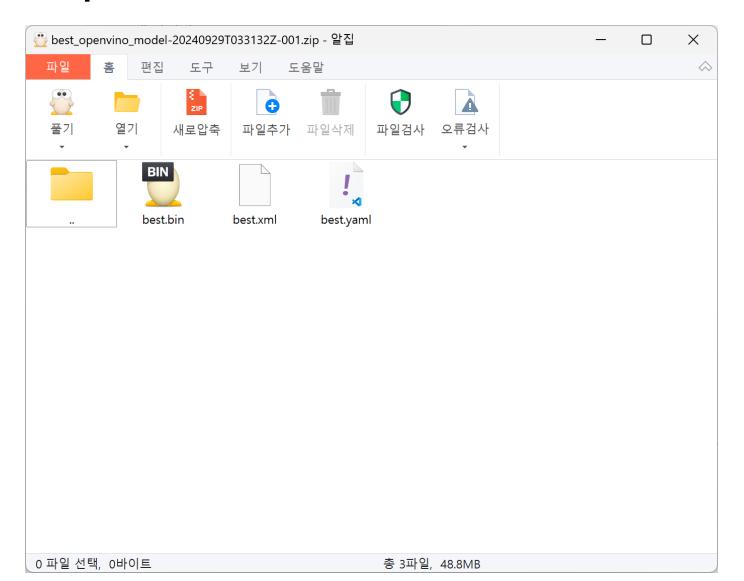
```
# specify input and output names for compatibility with yolov9 repo interface
ov_model.outputs[0].get_tensor().set_names({"output0"}))
ov_model.inputs[0].get_tensor().set_names({"images"})
ov.save_model(ov_model, ov_model_path)
# save metadata
yaml_save(ov_model_path.parent / weights.name.replace(".pt", ".yaml"), metadata)
else:
    metadata = yaml_load(ov_model_path.parent / weights.name.replace(".pt", ".yaml"))
```

/content/yolov9/models/experimental.py:243: FutureWarning: You are using `torch.load` with `weights_only=False` (the current defackpt = torch.load(attempt_download(w), map_location='cpu') # load
Fusing layers...
gelan-c summary: 387 layers, 25233256 parameters, 0 gradients, 101.8 GFLOPs
/content/yolov9/models/yolo.py:108: TracerWarning: Converting a tensor to a Python boolean might cause the trace to be incorrect.
elif self.dynamic or self.shape != shape:

OpenVINO 모델 다운로드



OpenVINO 모델 다운로드



cd C:/camera

conda create -n camera_env python=3.11

conda activate camera_env

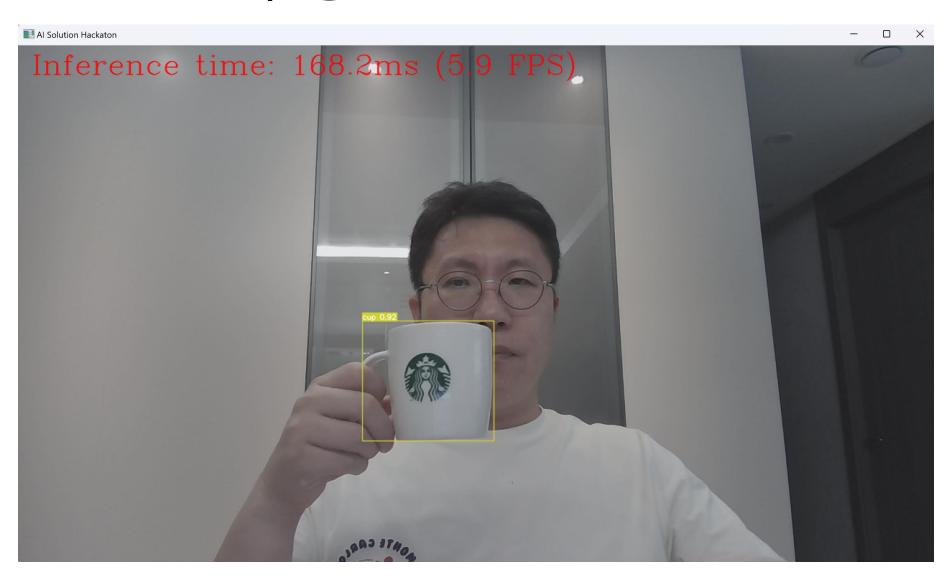
pip install -q "openvino>=2023.1.0"

pip install openvino-dev

pip install opency-python

python solution.py

```
# Webcam
VIDEO SOURCE = ∅
#file
#VIDEO SOURCE = 'test.mp4'
#VIDEO SOURCE =
'https://storage.openvinotoolkit.org/repositories/openvino_notebooks/data/data/video/people
.mp4'
source=VIDEO_SOURCE
flip=True
use popup=True
skip_first_frames=0
player = None
try:
    # Create a video player to play with target fps.
    player = VideoPlayer(source=source, flip=flip, fps=10,
skip_first_frames=skip_first_frames)
```



```
import os, glob
 from IPython.display import Image
 from google.colab import drive, userdata
 HOME = os.getcwd()
  YOLO = os.path.join(HOME, 'yolov9')
 print(HOME)
 print(YOLO)
 /content
 /content/volov9
 # 구글 드라이브 마운트
 drive.mount('/content/drive')
 Mounted at /content/drive
 pip install -q "openvino>=2023.3.0" "nncf>=2.8.1" "opencv-python" "seaborn" "pandas" "scikit-learn" "torch" "torchvision" "tqdm" --extra-index-url https://download.pytorch.org/whl/cpu
₹
  Preparing metadata (setup.py) ... done
      Preparing metadata (setup.py) ... done
         Building wheel for jstyleson (setup.py) ... done
  Building wheel for grapheme (setup.py) ... done
```

```
[6] !git clone https://github.com/WongKinYiu/yolov9.git
%cd yolov9
!pip install -r requirements.txt -q

Cloning into 'yolov9'...
remote: Enumerating objects: 781, done.
remote: Total 781 (delta 0), reused 0 (delta 0), pack-reused 781 (from 1)
Receiving objects: 100% (781/781), 3.27 MiB | 6.35 MiB/s, done.
Resolving deltas: 100% (331/331), done.
/content/yolov9/yolov9
```

▼ NNCF후처리 양자화 API로 모델 최적화

NNCF는 최소한의 성능 저하를 유지하면서 OpenVINO의 신경만 추론 최적화를 위한 알고리즘을 제공한다. YOLOv9을 최적화하기 위해서 후반 학습 모드로 8비트 양자화를 사용한다. 최적화 과정은 다음 단계로 구성된다.

- 양자화용 데이터셋 생성
- 최적화 모델을 얻기 위해 nncf.quantize 실행
- ov.save_model를 사용하여 OpenVINO IR 모델 저장

✔ 데이터셋 준비

기존 데이터셋을 재사용한다. yolov9 모델의 정확도를 평가하기 위해서 사용한다.

✔ 데이터셋 준비

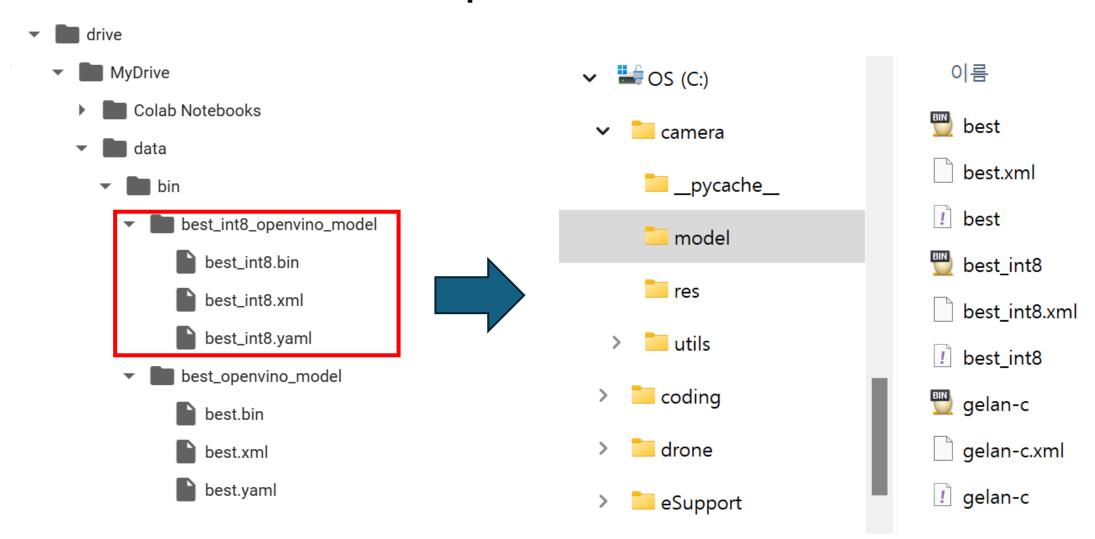
기존 데이터셋을 재사용한다. yolov9 모델의 정확도를 평가하기 위해서 사용한다.

```
from collections import namedtuple
import yaml
from utils.dataloaders import create_dataloader
from utils.general import colorstr
from pathlib import Path
 # read dataset config
DATA_CONFIG = '/content/drive/MyDrive/data/coco.yaml
with open(DATA CONFIG) as f:
    data = yaml.load(f, Loader=yaml.SafeLoader)
 # Dataloader
TASK = "val" # path to train/val/test images
Option = namedtuple("Options", ["single_cls"]) # imitation of commandline provided options for single class evaluation
opt = Option(False)
dataloader = create_dataloader(
    str(Path("/content/drive/MyDrive/data/coco") / data[TASK]).
    640.
    1.
    32.
    opt,
    pad=0.5.
    prefix=colorstr(f"{TASK}: "),
[0]
```

🚁 val: Scanning /content/drive/MyDrive/data/coco/val.cache... 2973 images, O backgrounds, O corrupt: 100%|

```
import numpy as np
import torch
from PIL import Image
from utils.augmentations import letterbox
def preprocess_image(img0: np.ndarray):
    Preprocess image according to YOLOv9 input requirements.
    Takes image in np.array format, resizes it to specific size using letterbox resize, converts color space from BGR (default in OpenCV) to RGB and changes data layout from HWC to CHW
    Parameters:
      imgO (np.ndarray): image for preprocessing
      img (np.ndarray): image after preprocessing
      imgO (np.ndarray): original image
    # resize
    img = letterbox(img0, auto=False)[0]
    # Convert
    img = img.transpose(2, 0, 1)
    img = np.ascontiguousarray(img)
    return img. imgO
def prepare_input_tensor(image: np.ndarray):
    Converts preprocessed image to tensor format according to YOLOv9 input requirements.
    Takes image in np.array format with unit8 data in [0, 255] range and converts it to torch. Tensor object with float data in [0, 1] range
    Parameters:
      image (np.ndarray): image for conversion to tensor
      input_tensor (torch.Tensor): float tensor ready to use for YOLOv9 inference
    input tensor = image.astvpe(np.float32) # uint8 to fp16/32
    input_tensor /= 255.0 # 0 - 255 to 0.0 - 1.0
    if input_tensor.ndim == 3:
        input_tensor = np.expand_dims(input_tensor, 0)
    return input_tensor
```

```
✓ [13] import nncf
         def transform_fn(data_item):
             Quantization transform function. Extracts and preprocess input data from dataloader item for quantization
            Parameters:
               data_item: Tuple with data item produced by DataLoader during iteration
            Returns:
                 input_tensor: Input data for quantization
             img = data_item[0].numpy()
             input_tensor = prepare_input_tensor(img)
             return input_tensor
         quantization_dataset = nncf.Dataset(dataloader, transform_fn)
        import openvino as ov
         from utils.general import yaml_save, yaml_load
         MODEL_DIR = Path("/content/drive/MyDrive/data/bin/")
         weights = MODEL DIR / "best.pt"
         ov_int8_model_path = MODEL_DIR / weights.name.replace(".pt", "_int8_openvino_model") / weights.name.replace(".pt", "_int8.xml")
         ov_model_path = MODEL_DIR / weights.name.replace(".pt", "_openvino_model") / weights.name.replace(".pt", ".xml")
         core = ov.Core()
         # read converted model
         ov_model = core.read_model(ov_model_path)
         metadata = yaml_load("/content/drive/MyDrive/data/bin/best_openvino_model/best.yaml")
         NAMES = metadata["names"]
         if not ov_int8_model_path.exists():
             quantized model = nncf.quantize(ov model, quantization dataset, preset=nncf.QuantizationPreset.MIXED)
            ov.save_model(quantized_model, ov_int8_model_path)
             yaml_save(ov_int8_model_path.parent / weights.name.replace(".pt", "_int8.yam|"), metadata)
    → Statistics collection
                                                                                               - 100% 300/300 • 0:26:34 • 0:00:00
                                                                                               - 100% 138/138 • 0:00:20 • 0:00:00
         Applying Fast Bias correction .
```



cd C:/camera

conda create –n camera_env python=3.11

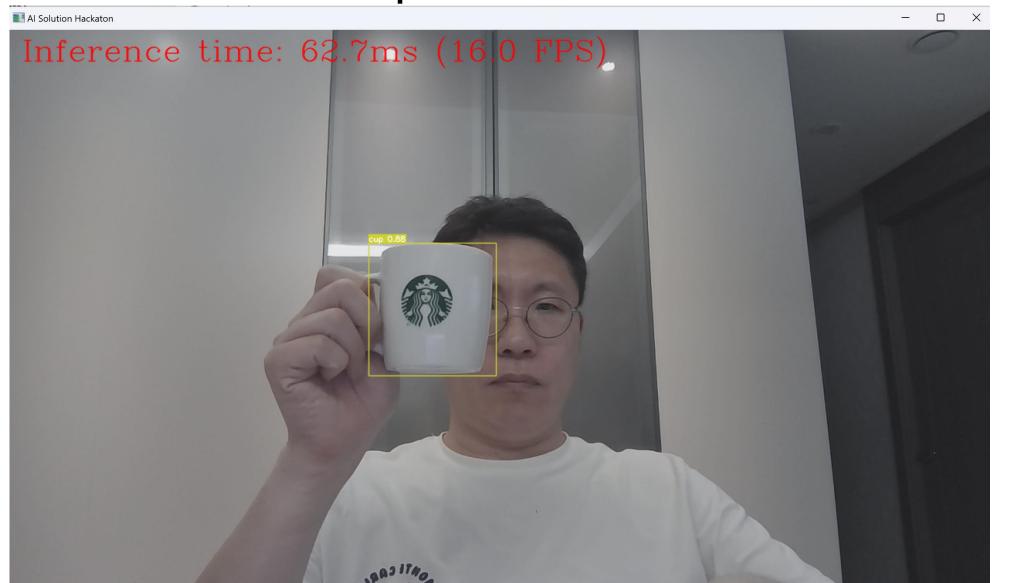
conda activate camera_env

pip install -q "openvino>=2023.1.0"

pip install openvino-dev

pip install opency-python

python quant.py



5.9FPS



16.0FPS