YOLOv9 QAT for TensorRT Detection / Segmentation



This repository contains an implementation of YOLOv9 with Quantization-Aware Training (QAT), specifically designed for deployment on platforms utilizing TensorRT for hardware-accelerated inference.

This implementation aims to provide an efficient, low-latency version of YOLOv9 for real-time detection applications.

If you do not intend to deploy your model using TensorRT, it is recommended not to proceed with this implementation.

- The files in this repository represent a patch that adds QAT functionality to the original YOLOv9 repository.
- This patch is intended to be applied to the main YOLOv9 repository to incorporate the ability to train with QAT.
- The implementation is optimized to work efficiently with TensorRT, an inference library that leverages hardware acceleration to enhance inference performance.
- Users interested in implementing object detection using YOLOv9 with QAT on TensorRT platforms can benefit from this repository as it provides a ready-to-use solution.

We use TensorRT's pytorch quntization tool to finetune training QAT yolov9 from the pre-trained weight, then export the model to onnx and deploy it with TensorRT. The accuray and performance can be found in below table.

For those who are not familiar with QAT, I highly recommend watching this video:

Quantization explained with PyTorch - Post-Training Quantization, Quantization-Aware Training

Getting started (工欲善其事,必先利其器)

For getting started, needs some steps.

Git Download

Git

Settings: git_command

Download Archive

yolov9

git clone https://github.com/WongKinYiu/yolov9.git

git status

git pull

yolov9-qat

git clone https://github.com/levipereira/yolov9-qat.git

git status

git pull

CUDA / cudnn Info

Check the info and choose the best fit to your device.

nvidia-smi

				CUDA
-3.11	Clang 16.0.0	Bazel 6.1.0	8.7	11.8
-3.11	Clang 16.0.0	Bazel 5.3.0	8.6	11.8
-3.11	GCC 9.3.1	Bazel 5.3.0	8.6	11.8
-3.10	GCC 9.3.1	Bazel 5.3.0	8.1	11.2
-3.10	GCC 9.3.1	Bazel 5.1.1	8.1	11.2
-3.10	GCC 9.3.1	Bazel 5.0.0	8.1	11.2
-3.10	GCC 7.3.1	Bazel 4.2.1	8.1	11.2
-3.9	GCC 7.3.1	Bazel 3.7.2	8.1	11.2
-3.9	GCC 7.3.1	Bazel 3.7.2	8.1	11.2
-3.9	GCC 7.3.1	Bazel 3.7.2	8.1	11.2
-3.8	GCC 7.3.1	Bazel 3.1.0	8.0	11.0
-3.8	GCC 7.3.1	Bazel 3.1.0	7.6	10.1
-3.8	GCC 7.3.1	Bazel 2.0.0	7.6	10.1
	3.11 3.11 3.10 3.10 3.10 3.10 3.9 3.9 3.9 3.9 3.8	GCC 9.3.1 GCC 7.3.1 GCC 7.3.1	GCC 9.3.1 Bazel 5.3.0 GCC 9.3.1 Bazel 5.1.1 GCC 9.3.1 Bazel 5.0.0 GCC 7.3.1 Bazel 4.2.1 GCC 7.3.1 Bazel 3.7.2 GCC 7.3.1 Bazel 3.7.2	GCC 9.3.1 Bazel 5.3.0 8.6 GCC 9.3.1 Bazel 5.3.0 8.6 GCC 9.3.1 Bazel 5.3.0 8.1 GCC 9.3.1 Bazel 5.3.0 8.1 GCC 9.3.1 Bazel 5.1.1 8.1 GCC 9.3.1 Bazel 5.0.0 8.1 GCC 9.3.1 Bazel 4.2.1 8.1 GCC 7.3.1 Bazel 3.7.2 8.1

Version	Python version	Compiler	Build tools	cuDNN	CUDA
tensorflow-2.1.0	2.7, 3.3-3.7	GCC 7.3.1	Bazel 0.27.1	7.6	10.1
tensorflow-2.0.0	2.7, 3.3-3.7	GCC 7.3.1	Bazel 0.26.1	7.4	10.0

CUDA cudnn pytorch TensorRT

Implementation Environment

Windows11 + WSL_Ubuntu-20.04(LTS)

Windows11 / WSL_Ubuntu-20.04-LTS

Windows11

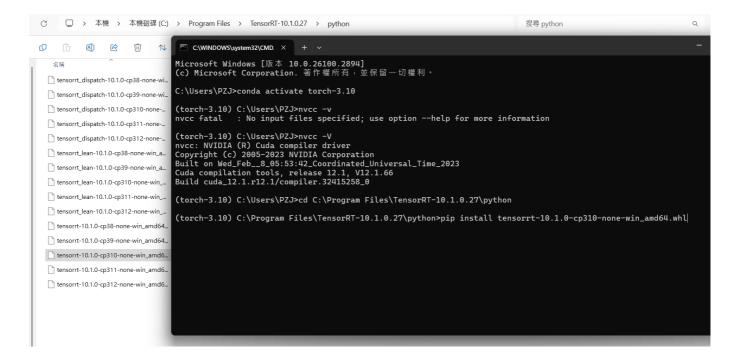
- Environment
 - o python 3.10.16
 - o CUDA 12.1
 - o cudnn 8.8
 - o pytorch

pip3 install torch torchvision torchaudio --index-url https://download.pytorch.org/whl/cu121

o TensorRT 10.1.0

TensorRT Installation Guide

- 1.Extract the downloaded files.
- 2.Set the necessary environment variables in the installation directory.
- 3.Add the bin and lib folders to the PATH.
- 4.Python_Building



WSL_Ubuntu-20.04-LTS

- Environment
 - o python 3.10.16
 - CUDA 11.8
 - o cudnn 8.6
 - pytorch

```
pip3 install torch torchvision torchaudio --index-url
https://download.pytorch.org/whl/cu118
```

TensorRT 10.1.0

TensorRT Installation Guide

- Main System Debian Installation
- Your Executing Environment (ex: torch-3.10)
- 1. NVIDIA TensorRT 10.x Download click the NVIDIA TensorRT License Agreement and find the archive that best fit your OS and devices.

Make sure you have activated the environment

Follow these steps to install TensorRT on WSL Ubuntu 20.04.

2. Install the Local TensorRT Repository

First, navigate to the directory where the TensorRT .deb package is located and install it using dpkg:

```
cd /mnt/c/Users/PZJ/Downloads
sudo dpkg -i nv-tensorrt-local-repo-ubuntu2004-10.1.0-cuda-11.8_1.0-1_amd64.deb
```

3. Add GPG Key and Update Package List

After installing the local repository package, copy the GPG key to the appropriate directory and update the package list:

```
sudo cp /var/nv-tensorrt-local-repo-ubuntu2004-10.1.0-cuda-11.8/*.pub
/usr/share/keyrings/
sudo apt-get update
```

4. Install TensorRT and Dependencies

To install TensorRT and its necessary dependencies, run the following commands:

```
sudo apt-get install -y tensorrt
sudo apt-get install -y python3-libnvinfer-dev
sudo apt-get install -y uff-converter-tf
sudo apt-get install -y onnx-graphsurgeon
```

5. Verify Installation

To check if TensorRT has been successfully installed, use the following command:

```
dpkg -1 | grep TensorRT
```

6. Test TensorRT Execution

To verify that TensorRT is functioning correctly, you can run an inference test using trtexec with an ONNX model:

```
trtexec --onnx=<your_model>.onnx
```

Notice

Please make sure that the TensorRT version built into the system is consistent with the version installed in Python.

Make sure environment variables are set correctly.

```
nano ~/.bashrc
```

Edit in nano

```
source ~/.bashrc
```

```
# <<< conda initialize <<<

# export PATH="/mnt/c/Users/PZJ/anaconda3/bin:$PATH" # commented out by conda initialize
export PATH=/usr/local/cuda-11.8/bin:$PATH
export LD_LIBRARY_PATH=/usr/local/cuda-11.8/lib64:$LD_LIBRARY_PATH

export PATH=/usr/src/tensorrt/bin:$PATH
export LD_LIBRARY_PATH=/usr/src/tensorrt/lib:$LD_LIBRARY_PATH

# for libnvinfer_plugin.so.10
export LD_LIBRARY_PATH=/home/user_pzj/anaconda3/envs/torch-3.10/lib/python3.10/site-packages/tensorrt_libs:$LD_LIBRARY_PATH

export PYTHONPATH=/home/user_pzj/anaconda3/envs/torch-3.10/lib/python3.10/site-packages:$PYTHONPATH</pre>
```

Verify the Version

```
(torch-3.10) user
                                                 /PU:/mnt/c/Users/PZJ$ dpkg-query -W tensorrt
tensorrt
                      10.1.0.27-1+cuda11.8
(torch-3.10) user
                                                  PU:/mnt/c/Users/PZJ$ dpkg-query -W "*nvinfer*"
libnvinfer-bin 10.1.0.27-1+cuda11.8
libnvinfer-dev 10.1.0.27-1+cuda11.8
libnvinfer-dev-cross-amd64
libnvinfer-dispatch-dev 10.1.0.27-1+cuda11.8
libnvinfer-dispatch-dev-cross-amd64
libnvinfer-dispatch10 10.1.0.27-1+cuda11.8
libnvinfer-doc
libnvinfer-headers-dev 10.1.0.27-1+cuda11.8
libnvinfer-headers-plugin-dev 10.1.0.27-1+cuda11.8
libnvinfer-lean-dev 10.1.0.27-1+cuda11.8
libnvinfer-lean-dev-cross-amd64
                                 10.1.0.27-1+cuda11.8
10.1.0.27-1+cuda11.8
libnvinfer-lean10
libnvinfer-plugin-dev
libnvinfer-plugin-dev-cross-amd64
libnvinfer-plugin10 10.1.0.27-1+cuda11.8
libnvinfer-samples 10.1.0.27-1+cuda11.8
libnvinfer-vc-plugin-dev
                                           10.1.0.27-1+cuda11.8
libnvinfer-vc-plugin-dev-cross-amd64
libnvinfer-vc-plugin10 10.1.0.27-1+cuda11.8
libnvinfer10 10.1.0.27-1+cuda11.8
libnvinfer10
python3-libnvinfer 10.1.0.27-1+cuda11.8
python3-libnvinfer-dev 10.1.0.27-1+cuda11.8
python3-libnvinfer-dispatch 10.1.0.27-1+cuda11.8
python3-libnvinfer-lean 10.1.0.27-1+cuda11.8
(torch-3.10) user_pzj@DESKTOP-20USVPU:/mnt/c/Users/PZJ$ pip show tensorrt
Name: tensorrt
Version: 10.1.0
Summary: TensorRT Metapackage
Home-page: https://developer.nvidia.com/tensorrt
Author: NVIDIA Corporation
Author-email:
License: Proprietary
Location: /home/user_pzj/anaconda3/envs/torch-3.10/lib/python3.10/site-packages
Requires: tensorrt-cu12
Required-by: nvidia-tensorrt
(torch-3.10) user_pzj@DESKTOP-20USVPU:/mnt/c/Users/PZJ$ python -c "import tensorrt as trt; print(trt.__version__)"
```

PS: The command python -c "import tensorrt as trt; print(trt.__version__)" -> (10.8.0.43) version here refers to the tensorrt-cu12 version, not the real Python TensorRT version. Among them tensorrt-cu12 is the underlying C++ package.

```
If didn't obey may cause: Attributeerror: 'nonetype' object has no attribute 
'num_io_tensors'
```

Other Testing

Verify CUDA

```
nvcc -V
```

or

deviceQuery

```
Device 9: "MYIDIA Gaforce RTX MBSD Lastop GDU"

CUDA Driver Version / Duntime Version

CUDA Capability Rajor/Rinor version number:

619 MapSMtoCores for SM 8.9 is undefined. Default to use 128 Cores/SM

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Memory Clock rate:

7225 Mhz

7226 Maximum Layered 2D Texture Size, (num) layers

7226 Size;

7236 Maximum Layered 2D Texture size, (num) layers

724 June 10 J
```

Verify cudnn

```
cat /usr/include/cudnn_version.h | grep CUDNN_MAJOR -A 2
```

or

cat /usr/include/cudnn_version.h | grep CUDNN_MAJOR -A 2

```
(torch-3.10) user_pzj@DESKTOP-20USVPU:~$ cat /usr/include/cudnn_version.h | grep CUDNN_MAJOR -A 2
#define CUDNN_MAJOR 8
#define CUDNN_MINOR 6
#define CUDNN_PATCHLEVEL 0
--
#define CUDNN_VERSION (CUDNN_MAJOR * 1000 + CUDNN_MINOR * 100 + CUDNN_PATCHLEVEL)
/* cannot use constexpr here since this is a C-only file */
```