Execute_QAT

Execute Record

Record

Steps：

Train the Model Using Training Session:

Data preparation

```
bash scripts/get_coco.sh
```

- Download MS COCO dataset images (train, val, test) and labels. If you have previously used a different version of YOLO, we strongly recommend that you delete `train2017.cache` and `val2017.cache` files, and redownload labels

Single GPU training

```
# train yolov9 models
!python train_dual.py --workers 8 --device 0 --batch 16 --data data/coco.yaml --
img 640 --cfg models/detect/yolov9-c.yaml --weights  yolov9-c.pt --name gelan-c --
hyp hyp.scratch-high.yaml --min-items 0 --epochs 100 --close-mosaic 15 --patience
10

# train gelan models
!python train.py --workers 8 --device 0 --batch 16 --data data/coco.yaml --img 640
--cfg models/detect/gelan-c.yaml --weights  gelan-c.pt --name gelan-c --hyp
hyp.scratch-high.yaml --min-items 0 --epochs 100 --close-mosaic 15 --patience 10
```

Multiple GPU training

```
# train yolov9 models
!python -m torch.distributed.launch --nproc_per_node 8 --master_port 9527
train_dual.py --workers 8 --device 0,1,2,3,4,5,6,7 --sync-bn --batch 128 --data
data/coco.yaml --img 640 --cfg models/detect/yolov9-c.yaml --weights '' --name
yolov9-c --hyp hyp.scratch-high.yaml --min-items 0 --epochs 500 --close-mosaic 15

# train gelan models
!python -m torch.distributed.launch --nproc_per_node 4 --master_port 9527 train.py
--workers 8 --device 0,1,2,3 --sync-bn --batch 128 --data data/coco.yaml --img 640
--cfg models/detect/gelan-c.yaml --weights '' --name gelan-c --hyp hyp.scratch-
high.yaml --min-items 0 --epochs 500 --close-mosaic 15
```

Reparameterize the Model reparameterization.py:

ex：FP32 -> FP16

Proceed with Quantization

```
!python qat.py quantize  --weights "C:\Users\PZJ\Desktop\yolov9\yolov9-
main_measure\yolov9-main\yolov9-c7-converted.pt" --data data/coco.yaml --hyp
./data/hyps/hyp.scratch-high.yaml --name qat_yolov9_reparameterize  --exist-ok
```

Eval Pytorch / Eval TensorRT (Will generate the val_trt file if succeed)：

Evaluate using Pytorch

```
!python qat.py eval --weights "C:\Users\PZJ\Desktop\yolov9\yolov9-
main_measure\yolov9-main\yolov9-c7-converted.pt"  --name
eval_qat_yolov9_reparameterize
```

Evaluate using TensorRT

Make sure to generate the .onnx through Export ONNX first !!

```
./scripts/val_trt.sh
runs/qat/qat_yolov9_reparameterize_SiLU/weights/qat_best_yolov9-c7-converted.pt
data/coco.yaml 640
```

Generate TensoRT Profiling and SVG image

```
./scripts/val_trt.sh
runs/qat/qat_yolov9_reparameterize_SiLU/weights/qat_best_yolov9-c7-converted.pt
data/coco.yaml 640 --generate-graph
```

Export ONNX

Export ONNX Model without End2End

```
!python export_qat.py --weights "C:\Users\PZJ\Desktop\yolov9\yolov9-
main_measure\yolov9-
main\runs\qat\qat_yolov9_reparameterize_SiLU\weights\qat_best_yolov9-c7-
converted.pt" --include onnx --dynamic --simplify --inplace
```

Export ONNX Model End2End

```
!python export_qat.py --weights "C:\Users\PZJ\Desktop\yolov9\yolov9-
main_measure\yolov9-
main\runs\qat\qat_yolov9_reparameterize_SiLU\weights\qat_best_yolov9-c7-
converted.pt" --include onnx_end2end
```

Deployment with Tensorrt

```
 /usr/src/tensorrt/bin/trtexec \
  --onnx=runs/qat/qat_yolov9_reparameterize_SiLU/weights/qat_best_yolov9-c7-
converted.onnx \
  --int8 --fp16  \
  --useCudaGraph \
  --minShapes=images:1x3x640x640 \
  --optShapes=images:4x3x640x640 \
  --maxShapes=images:8x3x640x640 \
  --saveEngine=runs/qat/qat_yolov9_reparameterize_SiLU/weights/qat_best_yolov9-c7-
converted.engine
```

Benchmark

Note: To test FP16 Models (such as Origin) remove flag `--int8`

```
# Set variable batch_size  and model_path_no_ext
# Mixed Precision
export batch_size=4
export
filepath_no_ext=runs/qat/qat_yolov9_reparameterize_SiLU/weights/qat_best_yolov9-
c7-converted
trtexec \
    --onnx=${filepath_no_ext}.onnx \
    --fp16 \
    --int8 \
    --saveEngine=${filepath_no_ext}_bs4_MP_SiLU.engine \
    --timingCacheFile=${filepath_no_ext}.engine.timing.cache \
    --warmUp=500 \
    --duration=10  \
    --useCudaGraph \
    --useSpinWait \
    --noDataTransfers \
    --minShapes=images:1x3x640x640 \
    --optShapes=images:${batch_size}x3x640x640 \
    --maxShapes=images:${batch_size}x3x640x640
```

Use Timing Comparison

功能	`val_trt.sh` 指令	`trtexec` 指令
模型輸入	ONNX 模型 (`qat_best_best.onnx`)	ONNX 模型 (`qat_best_yolov9-converted.onnx`)
功能	驗證流程，可能包含推理、效能測試、精度計算	轉換模型為 TensorRT 引擎，並進行效能測試
數據集設定	使用 `data/coco.yaml`	不涉及數據集驗證
動態輸入形狀	可能默認封裝設定	明確指定 `minShapes`, `optShapes`, `maxShapes`
精度選項	未顯式提及 FP16/INT8 模式	明確指定 `--fp16`, `--int8`
額外功能	生成計算圖 (`--generate-graph`)	無此功能
執行細節	封裝於腳本中，細節未展開	直接執行 `trtexec`，所有參數顯式定義

Result

- SiLU_MP
- SiLU_INT8
- SiLU_INT8