4 yolov5 model training

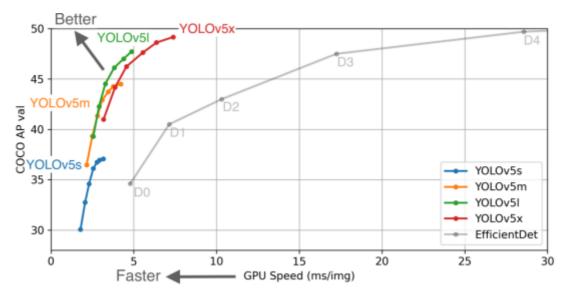
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4.1 yolov5 introduction

In February 2020, Joseph Redmon, the father of YOLO, announced his withdrawal from the field of computer vision research, YOLOv4 was released on April 23, 2020, and YOLOv5 was released on June 10, 2020. The developers of YOLOv5 claim that YOLOv5 can achieve fast detection at 140 FPS on Tesla P100; while the detection speed of YOLOv4 is 50 FPS. Not only that, but the size of YOLOv5 is only 27 MB. And YOLOv4 is 244 MB. YOLOv5 is nearly 90% smaller than YOLOv4. In terms of accuracy, YOLOv5 is comparable to YOLOv4.

As a result, YOLOv5 is very fast, has a very lightweight model size, and is comparable in accuracy to the YOLOv4 benchmark.

The algorithm performance test chart of the author of Yolov5:



Video test official case

```
cd ~/software/yolov5-5.0
python3 detection_video.py
```



4.2 folder structure

```
data
    argoverse_hd.yam1
    ├─ coco128.yaml
    ├─ coco.yaml
     — garbage
       ├─ Get_garbageData .py the py file of the corresponding
dataset(need more)
      ├— image
                                  # Save the original image(target
image to be recognized) folder
                                   # Test input image image folder
├─ texture
                                   # Folder to store background
  images(needs more)
# The image folder where the dataset
├─ images
is stored
          ├─ labels
                                  # label folder for storing datasets
garbage.yaml
     hyp _.fine tune.yaml
    hyp.scratch.yam1
     — images
                                  # Test input image image folder
       ├─ bus.jpg
       └─ zidane.jpg
```

```
- get_argoverse_hd.sh
                                       # Download dataset script
            get_coco.sh
             get_voc.sh
    └─ voc.yaml
                                        detect files
├─ detect # .py
  - Dockerfile
 — hubconf.py
 LICENSE
 - models
                                      # model configuration file
    ├─ yolo.py
    ├─ yolov5m.yaml
    ─ yolov5s.yaml
    yolov5x.yam1
    README.md
    requirements # .txt
                                       Environment requirements
    runs
    ├─ detect
        ∟— ехр
                                       # Test output image path folder
            ─ bus.jpg
            └─ train
        <u></u>— exp
           ├─ results.png
                                      # training result image
           results # .txt
                                         training log information
           test_batch0_labels.jpg
           test_batch0_pred.jpg
                                      # prediction image
           test_batch1_labels.jpg
           test_batch1_pred.jpg
           test_batch2_labels.jpg
            test_batch2_pred.jpg
                                     # training picture
           ├─ train_batch0.jpg
           ├─ train_batch1.jpg
             train_batch2.jpg
             weights
              ├─ best.pt
                                     # The best training model(generally
choose it)
                                           last trained model
               └─ last .pt The
                                        test file
— test # .py
                                        training file
├─ train # .py
├─ tutorial.ipynb
                                     # tutorial
├─ utils
└─ weights
                                     # The folder where the weight
files(pre-trained models) are stored
   download_weights.sh
   └─ yolov5s.pt
```

Note: There are verification pictures in the runs/exp file. After training an eTOCh, go and see to verify whether the pictures are correct. If they are not correct, adjust them quickly. Otherwise, you have to retrain after training all the adjustments.

label format

```
label x y w h
```

4.3 Environmental requirements

The factory image is already configured, no need to install

```
# pip install -r requirements.txt
# base -----
matplotlib > = 3.2.2
numpy > = 1.18.5
opencv - python > = 4.1.2
Pillow
PyYAML > = 5.3.1
scipy > = 1.4.1
torch > = 1.7.0
torchvision > = 0.8.1
tqdm > = 4.41.0
imgaug
# logging ------
tensorboard > = 2.4.1
# wandb
# plotting ------
seaborn > = 0.11.0
pandas
# export -----
# coremltools>=4.1
# onnx>=1.8.1
# scikit-learn==0.19.2 # for coreml quantization
# extras ------
thop # FLOPS computation
pycocotools > = 2.0 # COCO mAP
```

Installation example

```
pip install imgaug
```

4.4 Use process

- After configuring according to the above process
- Under the path of [data/garbage/texture], fill in some background images([more])
- Run the [Get_garbageData.py] file to get the dataset(line 134 can modify the number of generated datasets [more])
- Run the [train.py] file to start training
- Run [detect.py] for image prediction

4.5 Custom training data set

4.5.1 Collecting datasets

First go to Baidu to download or other methods, in the [data/garbage/texture] path, fill in some background images([more])

Open the [Get_garbageData.py] file

```
sudo vim Get_garbageData.py
```

Modify the total number of generated datasets and fill in as required. [More], too few datasets will lead to suboptimal training results.

```
img_total = 10000
```

Run the [Get_garbageData.py] file to get the dataset

```
python3 Get_garbageData.py
```

4.5.2 make yaml file

For example garbage.yaml:

```
train: ./garbage/train/images/  # Training set picture path
val: ./garbage/train/images/  # Verification set image path (also
separate from training set)
nc: 16  # number of categories
names: [ "Zip_top_can", "Apple_core" ...] # category name
```

4.5.3 Modify train.py

```
parser.add_argument('--weights', type = str, default = './weights/yolov5s.pt',
help = 'initial weights path') # Line 458:
parser.add_argument('--data', type = str, default = './garbage/garbage.yaml',
help = 'data.yaml path') # Line 460: Custom training file
parser.add_argument('--eTOChs', type = int, default = 300) # Line 462: Customize
training eTOChs, how many rounds of training
parser.add_argument('--batch-size', type = int, default = 16, help = 'total
batch size for all GPUs') #
parser.add_argument('--img-size', nargs = '+', type = int, default = [ 640, 640
], help = '[train, test] image sizes') # image size
parser.add_argument('--device', default = 'cpu', help = 'cuda device, ie 0 or
0,1,2,3 or cpu') # line 474: select CPU or GPU
parser.add_argument('--project', default = 'runs/train', help = 'save to
project/name') # training result output folder
```

Other places according to your needs.

4.5.4 Modify the model configuration file

Modify the second line of the yaml file of the yolov5 neural network, and use which weight file to modify the corresponding yaml file.

Here we are using yolov5s.yaml, so just modify the second line of the models/yolov5s.yaml file

```
nc : 16 # number of classes
```

4.5.5 Modify detect.py

Similar to the place where the [train.py] file needs to be modified

```
parser.add_argument('--weights', nargs = '+', type = str, default =
'weights/yolov5s.pt', help = 'model.pt path(s)') # Line 151: Pretrained weights
parser.add_argument('--source', type = str, default = 'data/images', help =
'source') # file/folder, 0 for webcam # Line 152: Input prediction image
parser.add_argument('--img-size', type = int, default = 640, help = 'inference
size(pixels)') # line 153: image size
parser.add_argument('--device', default = 'cpu', help = 'cuda device, ie 0 or
0,1,2,3 or cpu') # line 156: select CPU or GPU
parser.add_argument('--project', default = 'runs/detect', help = 'save results
to project/name') # Line 165: Prediction results output folder
```

Other places according to your needs.

4.5.6 Train and predict

After training the model, the final model will be produced in the [runs/train] folder after training.

```
# To use the trained best.pt weight file, it needs to be copied and pasted into
the yolov5-5.0 folder, then run the following command:
# To use the trained best.pt weight file, it needs to be copied and pasted into
the yolov5-5.0 folder, then run the following command:
python3 train.py
```

```
Disposable_chopsticks
                               100
                                             11
           Egg_shell
                              100
                                             6
                                                          0
                                                                       0
                                                                                    0
          Apple_core
                              100
                                            16
                                                          0
     Watermelon_rind
                              100
                                            13
                                                          0
                                                                       0
           Fish_bone
                                             9
                              100
                                                          0
                                                                       0
     Expired_tablets
                                             8
                              100
                                                          0
                                                                       0
   Expired_cosmetics
                                                                                    0
                              100
                                             9
                                                          0
                                                                       0
      Used_batteries
                                                          0
                                                                       0
                              100
                                            13
             Syringe
                              100
                                            10
                                                          0
                                                                       0
 epochs completed in 0.208 hours.
Optimizer stripped from runs/train/exp/weights/last.pt, 14.4MB
Optimizer stripped from runs/train/exp/weights/best.pt, 14.4MB
```

Image prediction, modify the model path and input image, predict the image, and the result is in the [runs/detect] folder.

Default model detection
python3 detect.py



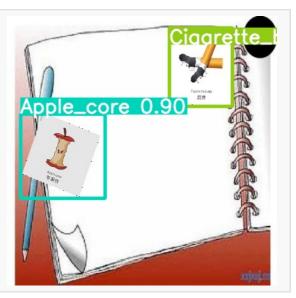


The detection effect below is from a model trained using 100 epochs and 10,000 images (since Jetson's performance is insufficient for training, the training was completed on Windows).

Copy and paste the trained best.pt weight file into the yolov5-5.0 folder, then run the following command.

python3 detect_elep.py





For real-time video monitoring, the model path needs to be modified.

python3 detection_video.py