

ATSS训练自己的数据集

[\[Repo\]](#) [\[paper\]](#)

1.Introduction

anchor-based和anchor-free最本质的区别是正负样本的定义 (**how to define positive and negative training samples**). ATSS (Adaptive Training Sample Selection)可以自动选择正负样本基于一些统计学的指标。ATSS提高了anchor based和anchor free的模型的识别精度, 使他们之间的gap变小。

2.Installation

对于环境的要求

```
Pytorch>=1.0
torchvision==0.2.1
cocoapi
yacs
matplotlib
GCC>=4.9 <6.0
python-opencv
```

安装:

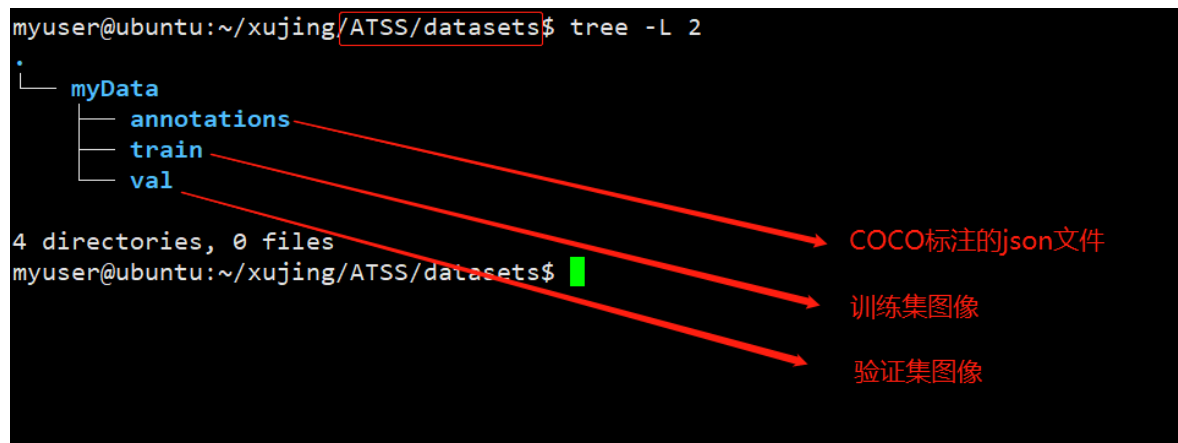
```
# 自行安装pytorch和torchvision
pip3 install ninja yacs cython matplotlib tqdm
# 安装cocoapi
git clone https://github.com/cocodataset/cocoapi.git
cd cocoapi/PythonAPI
python3 setup.py build_ext install
# atss
git clone https://github.com/sfzhang15/ATSS.git
cd ATSS

# cuda 9.0 ,9.2
sudo CUDA_HOST_COMPILER=/usr/bin/gcc-5 python3 setup.py build develop --no-deps
```

3.数据准备

将数据按照MS COCO的数据格式准备,在项目根目录下新建文件夹datasets/myData

```
myuser@ubuntu:~/xujing/ATSS/datasets$ tree -L 2
.
├── myData
│   ├── annotations
│   ├── train
│   └── val
└── 4 directories, 0 files
myuser@ubuntu:~/xujing/ATSS/datasets$
```



修改项目根目录下ATSS\atss_core\config\paths_catalog.py文件

```
class DatasetCatalog(object):
    DATA_DIR = "datasets"
    DATASETS = {

        "coco_2017_train": {
            "img_dir": "myData/train",
            "ann_file": "myData/annotations/instances_train.json"
        },
        "coco_2017_val": {
            "img_dir": "myData/val",
            "ann_file": "myData/annotations/instances_val.json"
        },
    }
```

4. 修改模型配置文件

修改模型配置文件 ATSS\configs\atss, 在该文件夹下新建文件夹比如 wei_score, 将 atss_dcnv2_X_101_64x4d_FPN_2x.yaml 配置文件拷贝到该文件夹并做如下修改 (部分参数可自行修改)

```
MODEL:
    META_ARCHITECTURE: "GeneralizedRCNN"
    WEIGHT: "catalog://ImageNetPretrained/FAIR/20171220/X-101-64x4d" #<-----预训练权重加载
    RPN_ONLY: True
    ATSS_ON: True
    BACKBONE:
        CONV_BODY: "R-101-FPN-RETINANET"
    RESNETS:
        STRIDE_IN_1X1: False
        BACKBONE_OUT_CHANNELS: 256
        NUM_GROUPS: 64
        WIDTH_PER_GROUP: 4
        STAGE_WITH_DCN: (False, False, True, True)
        WITH_MODULATED_DCN: True
        DEFORMABLE_GROUPS: 1
    RETINANET:
        USE_C5: False
    ATSS:
        ANCHOR_SIZES: (64, 128, 256, 512, 1024) # 8S
        ASPECT RATIOS: (1.0,)
        SCALES_PER_OCTAVE: 1
        USE_DCN_IN_TOWER: True
```

```

    POSITIVE_TYPE: 'ATSS' # how to select positives: ATSS (Ours) , SSC (FCOS), IoU
(RetinaNet)
    TOPK: 9 # topk for selecting candidate positive samples from each level
    REGRESSION_TYPE: 'BOX' # regressing from a 'BOX' or a 'POINT'
DATASETS:
    TRAIN: ("coco_2017_train",) #<-----与数据集对应
    TEST: ("coco_2017_val",)
INPUT:
    MIN_SIZE_RANGE_TRAIN: (640, 800)
    MAX_SIZE_TRAIN: 1333
    MIN_SIZE_TEST: 800
    MAX_SIZE_TEST: 1333
DATALOADER:
    SIZE_DIVISIBILITY: 32
SOLVER:
    BASE_LR: 0.01
    WEIGHT_DECAY: 0.0001
    STEPS: (120000, 160000)
    MAX_ITER: 180000
    IMS_PER_BATCH: 16 #<-----batch size可以修改
    WARMUP_METHOD: "constant"
TEST:
    BBOX_AUG:
        ENABLED: True #<-----多尺度测试
        VOTE: True
        VOTE_TH: 0.66
        MERGE_TYPE: "soft-vote"
        H_FLIP: True
        SCALES: (400, 500, 600, 640, 700, 900, 1000, 1100, 1200, 1300, 1400, 1800)
        SCALE_RANGES: [[96, 10000], [96, 10000], [64, 10000], [64, 10000], [64, 10000],
[0, 10000], [0, 10000], [0, 256], [0, 256], [0, 192], [0, 192], [0, 96]]
        MAX_SIZE: 3000
        SCALE_H_FLIP: True

```

5. 模型训练

```

python3 -m torch.distributed.launch \
    --nproc_per_node=8 \
    --master_port=$((RANDOM + 10000)) \
    tools/train_net.py \
    --config-file configs/atss/atss_R_50_FPN_1x.yaml \
    DATALOADER.NUM_WORKERS 2 \
    OUTPUT_DIR training_dir/atss_R_50_FPN_1x

```

- # 1. 如果你使用少的GPU, 可以更改--nproc_per_node, 指定GPU数量, 无需修改其他, batch size不依赖于该参数
- # 2. 如果你想改变batch size 则修改SOLVER.IMGS_PER_BATCH这个配置项中的参数
- # 3. 训练的模型江北保存在 OUTPUT_DIR
- # 4. 如果想修改训练的backbone, 可以修改--config-file

```
python3 -m torch.distributed.launch \
    --nproc_per_node=1 \
    tools/train_net.py \
    --config-file configs/atss/wei_score/atss_dcnv2_X_101_64x4d_FPN_2x.yaml \
    DATALOADER.NUM_WORKERS 2 \
    OUTPUT_DIR checkpoint/atss_dcnv2_X_101_64x4d_FPN_2x
```

```
python3 tools/train_net.py \
    --config-file configs/atss/wei_score/atss_dcnv2_X_101_64x4d_FPN_2x.yaml \
    DATALOADER.NUM_WORKERS 2 \
    OUTPUT_DIR checkpoint/atss_dcnv2_X_101_64x4d_FPN_2x\
    SOLVER.IMS_PER_BATCH 8
# 单卡 V100 32G能开到batch size 8
```

出现如下界面，则正常开始训练

```
loading annotations into memory...
Done (t=0.08s)
creating index...
index created!
2020-07-27 17:03:18,458 atss_core.trainer INFO: Start training
2020-07-27 17:05:36,820 atss_core.trainer INFO: eta: 14 days, 9:51:29 iter: 20 time: 7.5830 (6.9179) loss_reg: 0.9051 (0.9294) loss: 2.9748 (3.0321) data: 0.0172 (0.0618) 1
oss_cls: 1.2694 (1.4655) loss_centerness: 0.6278 (0.6373) lr: 0.003333 max mem: 22268
2020-07-27 17:07:47,250 atss_core.trainer INFO: eta: 13 days, 23:54:43 iter: 40 time: 6.6433 (6.7197) loss_reg: 0.8877 (0.9086) loss: 2.3512 (2.7355) data: 0.0180 (0.0400)
oss_cls: 0.8586 (1.1964) loss_centerness: 0.6214 (0.6306) lr: 0.003333 max mem: 22268
2020-07-27 17:09:58,640 atss_core.trainer INFO: eta: 14 days, 0:02:37 iter: 60 time: 7.5030 (6.7231) loss_reg: 0.8686 (0.8988) loss: 2.2371 (2.5795) data: 0.0184 (0.0328) 1
oss_cls: 0.7486 (1.0543) loss_centerness: 0.6176 (0.6265) lr: 0.003333 max mem: 22268
2020-07-27 17:12:18,619 atss_core.trainer INFO: eta: 14 days, 1:26:56 iter: 80 time: 7.3109 (6.7520) loss_reg: 0.8338 (0.8901) loss: 2.1622 (2.4890) data: 0.0195 (0.0294) 1
oss_cls: 0.6885 (0.9740) loss_centerness: 0.6209 (0.6249) lr: 0.003333 max mem: 22268
2020-07-27 17:14:31,368 atss_core.trainer INFO: eta: 14 days, 0:16:00 iter: 100 time: 6.9024 (6.7291) loss_reg: 0.8719 (0.8878) loss: 2.1715 (2.4288) data: 0.0213 (0.0278)
oss_cls: 0.6793 (0.9168) loss_centerness: 0.6211 (0.6242) lr: 0.003333 max mem: 22268
2020-07-27 17:16:46,744 atss_core.trainer INFO: eta: 14 days, 0:33:35 iter: 120 time: 6.9204 (6.7357) loss_reg: 0.8758 (0.8851) loss: 2.1523 (2.3822) data: 0.0217 (0.0268)
oss_cls: 0.6546 (0.8736) loss_centerness: 0.6185 (0.6235) lr: 0.003333 max mem: 22268
```

训练过程中GPU的占用情况

```
nvidia-smi -lms 200
```

```

+-----+
| Processes:                                     GPU Memory |
| GPU      PID    Type    Process name      Usage |
|=====|
| 0        28232   C       python3           26885MiB |
+-----+
Mon Jul 27 17:19:08 2020
+-----+
| NVIDIA-SMI 418.152.00    Driver Version: 418.152.00    CUDA Version: 10.1    |
|-----+-----+-----+-----+
| GPU  Name                Persistence-M| Bus-Id        Disp.A | Volatile Uncorr. ECC |
| Fan  Temp  Perf    Pwr:Usage/Cap|      Memory-Usage | GPU-Util  Compute M. |
|=====+=====+=====+=====|
|  0   Tesla V100-PCIE...    Off      | 00000000:00:06:0 Off |             0         |
| N/A   61C    P0      238W / 250W | 26896MiB / 32480MiB |      93%    Default  |
+-----+-----+-----+-----+
+-----+
| Processes:                                     GPU Memory |
| GPU      PID    Type    Process name      Usage |
|=====|
| 0        28232   C       python3           26885MiB |
+-----+

```

6. 模型推断

单张图像的推断结果的可视化,修改demo/predictor.py

```
CATEGORIES = [  
    "__background",  
    "QP",  
    "NY",  
    "QG",  
]
```

修改demo/atss_demo.py

```
# Copyright (c) Facebook, Inc. and its affiliates. All Rights Reserved.  
import argparse  
import cv2, os  
  
from atss_core.config import cfg  
from predictor import COCODemo  
  
import time  
  
def main():  
    parser = argparse.ArgumentParser(description="PyTorch Object Detection Webcam  
Demo")  
    parser.add_argument(  
        "--config-file",  
        default=" ../configs/atss/wei_score/atss_dcnv2_X_101_64x4d_FPN_2x.yaml", # <--  
        ---模型配置文件  
        metavar="FILE",  
        help="path to config file",  
    )  
    parser.add_argument(  
        "--weights",  
        default=" ../checkpoint/atss_dcnv2_X_101_64x4d_FPN_2x/model_0010000.pth",  
        #<-----训练模型地址  
        metavar="FILE",  
        help="path to the trained model",  
    )  
    parser.add_argument(  
        "--images-dir",  
        default=" ../datasets/myData/val", #<-----测试图像的路径  
        metavar="DIR",  
        help="path to demo images directory",  
    )  
    parser.add_argument(  
        "--min-image-size",  
        type=int,  
        default=800,  
        help="Smallest size of the image to feed to the model. "  
        "Model was trained with 800, which gives best results",  
    )  
    parser.add_argument(  
        "opts",  
        help="Modify model config options using the command-line",  
        default=None,  
        nargs=argparse.REMAINDER,  
    )  
  
    args = parser.parse_args()
```

```

# load config from file and command-line arguments
cfg.merge_from_file(args.config_file)
cfg.merge_from_list(args.opts)
cfg.MODEL.WEIGHT = args.weights

cfg.freeze()

# The following per-class thresholds are computed by maximizing
# per-class f-measure in their precision-recall curve.
# Please see compute_thresholds_for_classes() in coco_eval.py for details.

thresholds_for_classes = [
    0.5, 0.5, 0.5,
]

demo_im_names = os.listdir(args.images_dir)

# prepare object that handles inference plus adds predictions on top of image
coco_demo = COCODemo(
    cfg,
    confidence_thresholds_for_classes=thresholds_for_classes,
    min_image_size=args.min_image_size
)

for im_name in demo_im_names:
    img = cv2.imread(os.path.join(args.images_dir, im_name))
    if img is None:
        continue
    start_time = time.time()
    composite = coco_demo.run_on_opencv_image(img)
    print("{}\tinference time: {:.2f}s".format(im_name, time.time() - start_time))
    cv2.imwrite("../result/"+im_name, composite)
    # cv2.imshow(im_name, composite)
    # print("Press any keys to exit ...")
    # cv2.waitKey()
    # cv2.destroyAllWindows()

if __name__ == "__main__":
    main()

```

```
python3 atss_demo.py
```

```

38987686174811ea853300e04c510bc1.jpg    inference time: 0.21s
071bbe9c174811ea869900e04c510bc1.jpg    inference time: 0.25s
fa563722174711ea882300e04c510bc1.jpg    inference time: 0.25s
20200513_2019111816282724.jpg    inference time: 0.25s
20200513_264_20181112_two.jpg    inference time: 0.21s
42c06e10174811eab9b200e04c510bc1.jpg    inference time: 0.21s
20200513_1133_20181112_two.jpg    inference time: 0.21s
2a84fcac174811eaa51600e04c510bc1.jpg    inference time: 0.21s
f270f22e174711eaa22b00e04c510bc1.jpg    inference time: 0.21s
3b2db136174811eab7b300e04c510bc1.jpg    inference time: 0.21s
20200513_2019111812416486.jpg    inference time: 0.21s
1e53fbf4174811ea9e8b00e04c510bc1.jpg    inference time: 0.21s
21695cec174811eaba1100e04c510bc1.jpg    inference time: 0.21s
# 推断时间在V100上210ms

```


在V100上的推断过程中的显存占用:

Processes:					GPU	Memory Usage
GPU	PID	Type	Process name			
0	22790	C	python3			3203MiB
Tue Jul 28 14:17:58 2020						
NVIDIA-SMI 418.152.00 Driver Version: 418.152.00 CUDA Version: 10.1						
GPU	Name	Persistence-M	Bus-Id	Disp.A	Volatile Uncorr. ECC	
Fan	Temp	Perf	Pwr:Usage/Cap	Memory-Usage	GPU-Util	Compute M.
0	Tesla V100-PCIE...	Off	00000000:00:06.0	Off	0	
N/A	46C	P0	85W / 250W	3214MiB / 32480MiB	37%	Default
Processes:						
GPU	PID	Type	Process name		GPU	Memory Usage
0	22790	C	python3			3203MiB

7.部分测试结果展示

