

计算机视觉作业 7

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1. 作业要求

运行一个超分算法，训练一两个 Epoch，给出超分结果。

算法 Reference: <https://github.com/zzxvictor/License-super-resolution>

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本次作业



Github或者主页下载运行一个超分算法，获得结果试着训练一两个Epoch，给出超分结果

2. 实现过程

2.1 车牌增强超分算法

车牌增强是单图像超级分辨率（SISR）的更广泛领域的详细应用。

该项目受到了以下几种最先进 SR 模型的启发：

- Photo-Realistic Single Image Super-Resolution Using a Generative Adversarial Network
- Residual Dense Network for Image Super-Resolution
- ESRGAN: Enhanced Super-Resolution Generative Adversarial Networks
- Real-Time Single Image and Video Super-Resolution Using an Efficient Sub-Pixel Convolutional Neural Network

该项目使用的数据集来自：<https://github.com/detectRecog/CCPD>，可以直接下载使用。

2.2 安装项目依赖库

数据预处理：

- Dask >= 2.11.0
- PIL >= 6.2.2

训练与评估：

- tensorflow >= 2.1.0
- numpy >= 1.18.1
- matplotlib >= 3.1.3

2.3 训练过程

2.3.1 数据预处理

首先从 <https://github.com/detectRecog/CCPD> 下载数据集，然后运行 preprocess.py 文件进行数据预处理。方法：在终端输入 python preprocess.py 5

PATH_TO_UNZIPPED_DATA PATH_TO_OUTPUT_DIR 指令。

```
In [2]: from Utilities.io import DataLoader
        from Utilities.lossMetric import *
        from Utilities.trainVal import MinMaxGame
        from Models.RRDBNet import RRDBNet
        from Models.GAN import Discriminator
```

Load in the training dataset

I used the Chinese City Parking Dataset for this project. Please download the dataset from <https://github.com/detectRecog/CCPD>
Before loading the dataset, it is critical that you run the preprocessing script (preprocess.py) first!!! python preprocess.py 5
PATH_TO_UNZIPPED_DATA PATH_TO_OUTPUT_DIR

```
In [8]: import numpy as np
        import glob
        PATH = 'PATH_TO_OUTPUT_DIR/192_96' # only use images with shape 192 by 96 for training
        files = glob.glob(PATH + '/*.jpg') * 3 # data augmentation, same image with different brightness and contrast
        np.random.shuffle(files)
        train, val = files[:int(len(files)*0.8)], files[int(len(files)*0.8):]
        loader = DataLoader()
        trainData = DataLoader().load(train, batchSize=16)
        valData = DataLoader().load(val, batchSize=64)
```

2.3.2 训练 SR 模型

设定 Epoch=1, 尝试训练 SR 模型。

Training

```
In [9]: discriminator = Discriminator()
        extractor = buildExtractor()
        generator = RRDBNet(blockNum=10)

        Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg19/vgg19_weights_tf_dim_ordering_tf_kernels_notop.h5
        80134624/80134624 [=====] - 8s 0us/step
```

- It's a good idea to pretrain the generator model before the min-max game - Reference: <https://arxiv.org/abs/1701.00160>

```
In [10]: # a simple custom loss function that combines MAE loss with VGG loss, as defined in the SRGAN paper
        def contentLoss(y_true, y_pred):
            featurePred = extractor(y_pred)
            feature = extractor(y_true)
            mae = tf.reduce_mean(tfk.losses.mae(y_true, y_pred))
            return 0.1*tf.reduce_mean(tfk.losses.mse(featurePred, feature)) + mae

        optimizer = tfk.optimizers.Adam(learning_rate=1e-3)
        generator.compile(loss=contentLoss, optimizer=optimizer, metrics=[psnr, ssim])
        # epoch is set to 1 for demonstration purpose. In practice I found 20 is a good number
        # When the model reaches PSNR=20/ssim=0.65, we can start the min-max game
        history = generator.fit(x=trainData, validation_data=valData, epochs=1, steps_per_epoch=3, validation_steps=1)

        3/3 [=====] - 22s 8s/step - loss: 8.2637 - psnr: 2.1568 - ssim: 0.0106 - val_loss: 5.1731
        - val_psnr: 4.6175 - val_ssim: 0.0078
```

2.3.3 训练 GAN 增强模型

Generative adversarial network training

```
In [11]: # training parameter. epoch is set to 1 for demonstration
        # please train the network until it reaches snRatio ~ 22
        PARAMS = dict(lrGenerator = 1e-4,
                      lrDiscriminator = 1e-4,
                      epochs = 1,
                      stepsPerEpoch = 500,
                      valSteps = 100)
        game = MinMaxGame(generator, discriminator, extractor)
        log, valLog = game.train(trainData, valData, PARAMS)
        # ideally peak signal noise ratio(snRatio or psnr) should reach ~22
```

3. 实验结果

使用作者训练好的模型, 做超分车牌增强处理。

3.1 导入预训练模型

Load in the pretrained super-resolution model

```
In [3]: # pretrained rrdn network can be found in the Pretrained folder
        MODEL_PATH = 'Pretrained/rrdb'
        model = RRDBNet(blockNum=10)
        model.load_weights(MODEL_PATH)

        Out[3]: <tensorflow.python.training.tracking.util.CheckpointLoadStatus at 0x7ff601906040>
```

3.2 测试模型

Run plate enhancement

```
In [4]: for downSample, original in data.take(4):  
        yPred = model.predict(downSample)  
        painter.plot(downSample, original, yPred)
```

```
1/1 [=====] - 1s 679ms/step  
1/1 [=====] - 0s 65ms/step  
1/1 [=====] - 0s 65ms/step  
1/1 [=====] - 0s 64ms/step
```

