

ESRGAN: Enhanced Super-Resolution Generative Adversarial Networks

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1 Empirical Analysis

This paper proposed ESRGAN model that achieves consistently better perceptual quality than previous SR methods

1. Required dependencies: python==3.7, tensorflow-gpu==2.1.0, CUDA10.1, numpy, opencv-python, PyYAML, tqdm, Pillow

2. Dataset:

training data:DIV2k and Flickr2K where Low Resolution images for training : DIV2K_train_LR_

Ground-Truth Hight Resolution Images: DIV2K_train_HR

Test dataset: Set5, Set14, BSD100 Set14

Subimages newly created dataset for Interpolation:

DIV2K800_sub_bicLRx4 (LR), DIV2K800_sub (GT)

3.Hyperparameter setting:

batch_size: 16, input_size: 32, gt_size: 128, ch_size: 3, scale: 4,

4.Models: Models are trained in RGB channels, and the training dataset is augmented with random horizontal flips and 90-degree rotations.

2 Qualitative Results

2.1 Running rename.py file to rename image file in LR folder DIV2K_train_LR_bicubic

```
root@dgx-2:/workspace/data/esrgan-tf2/data# python rename.py  
Finished.
```


2.4 Running train_esrgan.py to train the ESRGAN model with the pretrain PSNR model 5 days on DGX-2

```
(base) root@dgx-2:/workspace/data/esrgan-tf2# python train_esrgan.py --cfg_path="./configs/esrgan.yaml" --gpu=0
Model: "RRDB_model"
```

Layer (type)	Output Shape	Param #	Connected to
input_image (InputLayer)	(None, 32, 32, 3, 0)		
conv_first (Conv2D)	(None, 32, 32, 64, 1792)		input_image[0][0]
RRDB_trunk (Sequential)	(None, 32, 32, 64, 16546752)		conv_first[0][0]
conv_trunk (Conv2D)	(None, 32, 32, 64, 36928)		RRDB_trunk[0][0]
tf_op_layer_Addv2 (Tensor)	(None, 32, 32, 6, 0)		conv_first[0][0] conv_trunk[0][0]
tf_op_layer_ResizeNearest (Tensor)	(None, 64, 64, 6, 0)		tf_op_layer_Addv2[0][0]
upconv_1 (Conv2D)	(None, 64, 64, 64, 36928)		tf_op_layer_ResizeNearestNe
tf_op_layer_ResizeNearest (Tensor)	(None, 128, 128, 0)		upconv_1[0][0]
upconv_2 (Conv2D)	(None, 128, 128, 36928)		tf_op_layer_ResizeNearestNe
conv_hr (Conv2D)	(None, 128, 128, 36928)		upconv_2[0][0]
conv_last (Conv2D)	(None, 128, 128, 1731)		conv_hr[0][0]

Total params: 16,697,987
Trainable params: 16,697,987
Non-trainable params: 0

```
Model: "Discriminator_VGG_128"
```

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	(None, 128, 128, 3)	0
linear2 (Dense)	(None, 1)	101

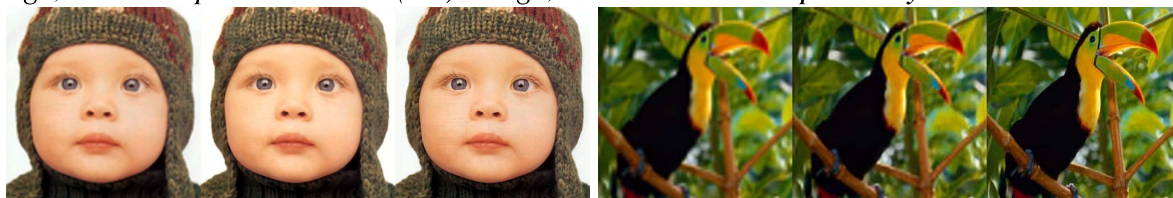
Total params: 14,505,161
Trainable params: 14,499,401
Non-trainable params: 5,760

```
10619 07:44:24.506086 140022443017984 utils.py:39] load train_dataset from /workspace/data/esrgan-tf2/data/Div2K800_sub_bin.tfrecord
[*] training from pretrain model ./checkpoints/psnr_pretrain.
Training [ ] 151/400000, loss_G=1.0536, loss_D=0.0019, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 152/400000, loss_G=1.3132, loss_D=0.1490, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 153/400000, loss_G=1.2330, loss_D=0.0431, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 154/400000, loss_G=1.1995, loss_D=0.1376, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 155/400000, loss_G=1.1946, loss_D=0.0075, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 156/400000, loss_G=1.0302, loss_D=0.0431, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 157/400000, loss_G=1.1923, loss_D=0.0943, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 158/400000, loss_G=1.1098, loss_D=0.0046, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
Training [ ] 159/400000, loss_G=1.2056, loss_D=0.0298, lr_G=1.0e-04, lr_D=1.0e-04 0.1 step/sec
```

2.5 Test and Result of PSNR model on Set5 and Set14. The corresponding images stored in Result folder.

```
(base) root@dgx-2:/workspace/data/esrgan-tf2# python test.py --cfg_path="./configs/psnr.yaml"
[*] load ckpt from ./checkpoints/psnr_pretrain/ckpt-1.
[*] Processing on set5 and set14, and write results
'set5_path' form /workspace/data/esrgan-tf2/data/set5
PSNR/SSIM
[head.png] Bic=32.01db/0.76, SR=31.56db/0.78
[bird.png] Bic=30.27db/0.87, SR=31.32db/0.89
[woman.png] Bic=26.44db/0.83, SR=27.44db/0.87
[baby.png] Bic=31.96db/0.85, SR=31.53db/0.88
[butterfly.png] Bic=22.25db/0.72, SR=23.98db/0.79
'set14_path' form /workspace/data/esrgan-tf2/data/set14
PSNR/SSIM
[barbara.png] Bic=25.19db/0.69, SR=25.40db/0.73
[monarch.png] Bic=27.60db/0.88, SR=29.11db/0.90
[face.png] Bic=31.98db/0.76, SR=31.47db/0.78
[pepper.png] Bic=29.38db/0.84, SR=29.10db/0.84
[man.png] Bic=25.74db/0.68, SR=26.17db/0.72
[baboon.png] Bic=22.06db/0.45, SR=22.09db/0.49
[foreman.png] Bic=27.65db/0.86, SR=27.24db/0.88
[coastguard.png] Bic=25.33db/0.52, SR=25.15db/0.56
[ppt3.png] Bic=21.76db/0.82, SR=23.10db/0.87
[bridge.png] Bic=24.38db/0.56, SR=24.46db/0.62
[lenna.png] Bic=29.67db/0.80, SR=30.17db/0.82
[zebra.png] Bic=24.15db/0.68, SR=25.73db/0.74
[flowers.png] Bic=25.85db/0.72, SR=26.53db/0.76
[comic.png] Bic=21.69db/0.59, SR=22.14db/0.65
[*] write the visual results in ./results/psnr_pretrain/
(base) root@dgx-2:/workspace/data/esrgan-tf2#
```

Result of Set5 using PSNR model, Three images showing Low Resolution Bicubic(Bic) image, PSNR Super Resolution(SR) image, Ground Truth respectively





Result of Set14 using PSNR model, Three images showing Low Resolution Bicubic(Bic) image, PSNR Super Resolution(SR) image, Ground Truth respectively

