```
1 from google.colab import drive
 2 drive.mount("/content/drive")
     Mounted at /content/drive
 1 import numpy as np
 2 import pandas as pd
 3 import os
 4 print(os.listdir("/content/drive/MyDrive/animated faces/animeface-character-dataset/data/"))
     ['face_0_421_57.png', 'face_0_149_83.png', 'face_0_343_119.png', 'face_0_421_23.png', 'face_0_119_15.png', 'face_13_303_75.png', 'face_1
1 import numpy as np
 2 import pandas as pd
 3 import os
 4 import time
 5 import tensorflow as tf
 6 import numpy as np
 7 import glob
8 from glob import glob
9 import datetime
10 import random
11 from PIL import Image
12 import matplotlib.pyplot as plt
13 import seaborn as sns
14 %matplotlib inline
15 import keras
16 from keras.layers import Input, Dense, Reshape, Flatten, Dropout
17 from keras.layers import BatchNormalization, Activation, ZeroPadding2D
18 from keras.layers import LeakyReLU
19 from keras.layers import UpSampling2D, Conv2D, Conv2DTranspose
20 from keras.models import Sequential, Model
21 from keras.optimizers import Adam
22 import warnings
23 warnings.filterwarnings("ignore")
1 IMAGE_SIZE = 64
 2 NOISE_SIZE = 100
3 LR D = 0.00004
 4 LR_G = 0.0004
 5 BATCH_SIZE = 32
 6 EPOCHS = 50
7 BETA1 = 0.5
8 WEIGHT_INIT_STDDEV = 0.02
 9 EPSILON = 0.00005
10 SAMPLES_TO_SHOW = 8
1 def get_generator(z=(NOISE_SIZE,)):
       # 4 x 4 x 512
 3
      input laver = Input(z)
      hid = Dense(4*4*512, activation='relu', name ="Dense")(input_layer)
 4
 5
      hid = LeakyReLU(alpha=0.2)(hid)
      hid = Reshape((4, 4, 512))(hid)
 6
 7
      hid = Conv2DTranspose(512, kernel_size=[5,5],
 8
                                  strides=[2,2],
 9
                                  padding="same",
10
                                  kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),name ="trans_conv1")(hid)
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch_trans_conv1")(hid)
11
12
      hid = LeakyReLU(alpha=0.2,name ="trans_conv1_out")(hid)
      hid = Conv2DTranspose(256, kernel_size=[5,5],
13
14
                                  strides=[2,2],
15
                                  padding="same"
16
                                  kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),name ="trans_conv2")(hid)
17
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch_trans_conv2")(hid)
      hid = LeakyReLU(alpha=0.2,name ="trans conv2 out")(hid)
18
       hid = Conv2DTranspose(128, kernel_size=[5,5],
19
20
                                  strides=[2,2],
21
                                  padding="same"
22
                                  kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),name ="trans_conv3")(hid)
23
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch_trans_conv3")(hid)
       hid = LeakyReLU(alpha=0.2,name ="trans_conv3_out")(hid)
24
      hid = Conv2DTranspose(64, kernel_size=[5,5],
```

```
26
                                  strides=[2,2],
27
                                  padding="same"
                                  kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),name ="trans_conv4")(hid)
28
29
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch_trans_conv4")(hid)
30
      hid = LeakyReLU(alpha=0.2,name ="trans_conv4_out")(hid)
31
      hid = Conv2DTranspose(3, kernel_size=[5,5],
32
                                  strides=[1,1],
33
                                  padding="same"
34
                                  kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),name ="logits")(hid)
35
      out = Activation("tanh", name ="out")(hid)
      model = Model(inputs=input_layer, outputs=out)
36
37
       model.summary()
38
      return model
1 def get_discriminator(input_shape=(IMAGE_SIZE, IMAGE_SIZE,3)):
       input_layer = Input(input_shape)
 2
 3
      hid = Conv2D(filters=32,
 4
                       kernel_size=[5,5],
 5
                       strides=[2,2],
 6
                       padding="same",
 7
                       kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),
                       name = "conv1")(input_layer)
 8
 9
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch_norm1")(hid)
10
      hid = LeakyReLU(alpha=0.2, name="conv1_out")(hid)
      hid = Conv2D(filters=64,
11
12
                           kernel_size=[5,5],
13
                           strides=[2,2],
14
                           padding="same"
                           kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),
15
                           name = "conv2")(hid)
16
17
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch_norm2")(hid)
18
      hid = LeakyReLU(alpha=0.2, name="conv2_out")(hid)
19
       hid = Conv2D(filters=128,
20
                       kernel_size=[5,5],
21
                       strides=[2,2],
22
                       padding="same",
23
                       kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),
24
                       name = "conv3")(hid)
25
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch_norm3")(hid)
      hid = LeakyReLU(alpha=0.2, name="conv3 out")(hid)
26
27
      hid = Conv2D(filters=256,
28
                       kernel size=[5,5],
29
                       strides=[1,1],
30
                       padding="same"
31
                       kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),
32
                       name = "conv4")(hid)
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch norm4")(hid)
33
       hid = LeakyReLU(alpha=0.2, name="conv4_out")(hid)
34
35
      hid = Conv2D(filters=512,
36
                       kernel_size=[5,5],
37
                       strides=[2,2],
38
                       padding="same"
39
                       kernel_initializer= keras.initializers.TruncatedNormal(stddev=WEIGHT_INIT_STDDEV),
40
                       name = "conv5")(hid)
41
      hid = BatchNormalization(momentum=0.9, epsilon=EPSILON, name="batch norm5")(hid)
42
      hid = LeakyReLU(alpha=0.2, name="conv5_out")(hid)
      hid = Flatten(name = "flatten")(hid)
43
       out = Dense(1, activation='sigmoid', name = "ligit")(hid)
44
45
      model = Model(inputs= input_layer, outputs=out)
46
      model.summary()
47
       return model
1 discriminator = get discriminator((IMAGE SIZE, IMAGE SIZE,3))
 2 discriminator.compile(loss='binary_crossentropy',optimizer=Adam(lr=LR_D, beta_1=BETA1),metrics=['accuracy'])
 3 discriminator.trainable = False
4 generator = get_generator((NOISE_SIZE,))
 5 gan_input = Input(shape=(NOISE_SIZE,))
 6 \times = generator(gan_input)
 7 gan_out = discriminator(x)
 8 gan = Model(gan_input, gan_out)
9 gan.summary()
10 gan.compile(loss='binary_crossentropy',optimizer=Adam(lr=LR_G, beta_1=BETA1))
11
```

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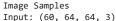
22 23

24

)

```
trans_conv2 (Conv2DTranspo (None, 16, 16, 256)
                                                         3277056
     batch_trans_conv2 (BatchNo (None, 16, 16, 256)
                                                         1024
     rmalization)
     trans_conv2_out (LeakyReLU (None, 16, 16, 256)
     trans_conv3 (Conv2DTranspo
                                (None, 32, 32, 128)
                                                         819328
     batch_trans_conv3 (BatchNo (None, 32, 32, 128)
                                                         512
     rmalization)
     trans_conv3_out (LeakyReLU (None, 32, 32, 128)
     trans conv4 (Conv2DTranspo
                                                         204864
                               (None, 64, 64, 64)
     batch_trans_conv4 (BatchNo (None, 64, 64, 64)
                                                         256
     rmalization)
     trans_conv4_out (LeakyReLU (None, 64, 64, 64)
     logits (Conv2DTranspose)
                                (None, 64, 64, 3)
                                                         4803
     out (Activation)
                                (None, 64, 64, 3)
    ______
    Total params: 11691395 (44.60 MB)
    Trainable params: 11689475 (44.59 MB)
    Non-trainable params: 1920 (7.50 KB)
    Model: "model 2"
     Layer (type)
                                Output Shape
                                                         Param #
     input_3 (InputLayer)
                                [(None, 100)]
     model_1 (Functional)
                                (None, 64, 64, 3)
                                                         11691395
     model (Functional)
                                (None, 1)
                                                         4367553
    _____
    Total params: 16058948 (61.26 MB)
    Trainable params: 11689475 (44.59 MB)
    Non-trainable params: 4369473 (16.67 MB)
    WARNING:absl:`lr` is deprecated in Keras optimizer, please use `learning_rate` or use the legacy optimizer, e.g.,tf.keras.optimizers
1 def show_samples(sample_images, name, epoch):
      figure, axes = plt.subplots(1, len(sample_images), figsize = (IMAGE_SIZE, IMAGE_SIZE))
      for index, axis in enumerate(axes):
          axis.axis('off')
          image_array = sample_images[index]
          axis.imshow(image_array)
          image = Image.fromarray(image_array)
      plt.show()
      plt.close()
10 def test(input_z, epoch):
      samples = generator.predict(input_z[:SAMPLES_TO_SHOW])
      sample_images = [((sample + 1.0) * 127.5).astype(np.uint8) for sample in samples]
      show_samples(sample_images, OUTPUT_DIR + "samples", epoch)
14 def summarize_epoch(d_losses, g_losses , data_shape, epoch, duration, input_z):
      minibatch_size = int(data_shape[0]//BATCH_SIZE)
      print("Epoch {}/{}".format(epoch, EPOCHS),
            "\nDuration: {:.5f}".format(duration),
            "\nD Loss: {:.5f}".format(np.mean(d_losses[-minibatch_size:])),
            "\nG Loss: {:.5f}".format(np.mean(g_losses[-minibatch_size:])))
      fig, ax = plt.subplots()
      plt.plot(d_losses, label='Discriminator', alpha=0.6)
      plt.plot(g_losses, label='Generator', alpha=0.6)
      plt.title("Losses")
      plt.legend()
      plt.savefig(OUTPUT_DIR + "losses_" + str(epoch) + ".png")
```

```
26
      plt.show()
27
      plt.close()
      test(input_z, epoch)
28
29 def get_batches(data):
30
      batches = []
      for i in range(int(data.shape[0]//BATCH_SIZE)):
31
           batch = data[i * BATCH_SIZE:(i + 1) * BATCH_SIZE]
32
           augmented_images = []
33
34
           for img in batch:
35
              image = Image.fromarray(img)
36
              if random.choice([True, False]):
37
                   image = image.transpose(Image.FLIP_LEFT_RIGHT)
38
               augmented_images.append(np.asarray(image))
39
           batch = np.asarray(augmented images)
40
           normalized_batch = (batch / 127.5) - 1.0
41
           batches.append(normalized_batch)
      return np.array(batches)
42
1 INPUT_DATA_DIR = "/content/drive/MyDrive/animated faces/animeface-character-dataset/data/"
2 OUTPUT_DIR =""
1 from PIL import Image
2 import re
3 exclude_img = []
4 exclude_img = [s + ".png" for s in exclude_img]
5 print("Image Samples")
6 input_images = np.asarray([np.asarray(Image.open(file).resize((IMAGE_SIZE, IMAGE_SIZE))) for file in glob(INPUT_DATA_DIR + '*') if file no
7 print ("Input: " + str(input_images.shape))
 8 np.random.shuffle(input_images)
9 sample_images = random.sample(list(input_images), SAMPLES_TO_SHOW)
10 show_samples(sample_images, OUTPUT_DIR + "inputs", 0)
```













```
1 print("Training Starts!")
 2 warnings.filterwarnings("ignore")
 3 d losses = []
 4 g_losses = []
 5 \text{ cum\_d\_loss} = 0
 6 \text{ cum}_gloss = 0
 7 for epoch in range(EPOCHS):
 8
       epoch += 1
9
       start_time = time.time()
10
       for batch_images in get_batches(input_images):
           noise_data = np.random.normal(0, 1, size=(BATCH_SIZE, NOISE_SIZE))
11
12
           generated_images = generator.predict(noise_data)
13
           noise_prop = 0.05
14
           real_labels = np.zeros((BATCH_SIZE, 1)) + np.random.uniform(low=0.0, high=0.1, size=(BATCH_SIZE, 1))
           flipped_idx = np.random.choice(np.arange(len(real_labels)), size=int(noise_prop*len(real_labels)))
15
           real_labels[flipped_idx] = 1 - real_labels[flipped_idx]
16
17
           d_loss_real = discriminator.train_on_batch(batch_images, real_labels)
18
           fake labels = np.ones((BATCH SIZE, 1)) - np.random.uniform(low=0.0, high=0.1, size=(BATCH SIZE, 1))
19
           flipped_idx = np.random.choice(np.arange(len(fake_labels))), size=int(noise_prop*len(fake_labels)))
           fake_labels[flipped_idx] = 1 - fake_labels[flipped_idx]
20
21
           d_loss_fake = discriminator.train_on_batch(generated_images, fake_labels)
           d_loss = 0.5 * np.add(d_loss_real, d_loss_fake)
22
23
           cum_d_loss += d_loss
24
           d_losses.append(d_loss[0])
25
           noise_data = np.random.normal(0, 1, size=(BATCH_SIZE, NOISE_SIZE))
26
           g_loss = gan.train_on_batch(noise_data, np.zeros((BATCH_SIZE, 1)))
27
           cum_g_loss += g_loss
28
           g_losses.append(g_loss)
29
       if epoch > 0 and epoch % 20 == 0:
           print("saving model")
30
31
           discriminator.save_weights("desc-simposon-model.h5-" + str(epoch))
32
           gan.save_weights("gan-simposon-model.h5-" + str(epoch))
       summarize epoch(d losses, g losses, input images.shape, epoch, time.time()-start time, noise data)
```

 \supseteq