img_gen

November 18, 2022

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[]: import tensorflow as tf
     gpus = tf.config.list_physical_devices('GPU')
     if gpus:
         tf.config.set_logical_device_configuration(gpus[0], [tf.config.
      →LogicalDeviceConfiguration(memory_limit=5292)])
     import keras
     import matplotlib.pyplot as plt
     import numpy as np
     from keras import layers, optimizers, losses, metrics, Model
     from PIL import Image
     from sklearn import preprocessing as pre
[]: # Construct the discriminator.
     def create D():
         input_img = keras.Input(shape=(256,256, 1))
         conv1 = layers.Conv2D(8, (2,2), (2,2), padding='valid',
                                     activation=layers.ReLU())(input_img)
         drop1 = layers.Dropout(rate=0.5)(conv1)
         down1 = layers.MaxPool2D(pool_size=(2,2), strides=(2,2),__
      ⇔padding="valid")(drop1)
         flat1 = layers.Flatten()(down1)
         score = layers.Dense(1, activation='sigmoid')(flat1)
         D: Model = Model(inputs=input_img, outputs=score)
         D.compile(optimizer=optimizers.Nadam(learning_rate=0.002), loss=losses.
      →BinaryCrossentropy(), metrics=metrics.MeanAbsoluteError())
         return D
[]: # Construct the generator.
     def create_G():
         random_input = keras.Input(shape=(100, ))
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B_Norm1 = layers.BatchNormalization()(Dense1)
         Relu1 = layers.LeakyReLU()(B_Norm1)
         reshape1 = layers.Reshape(target_shape=(16, 16, 1))(Relu1)
         DeConv1 = layers.Conv2DTranspose(filters=4, kernel_size=(2, 2),
      ⇔strides=(1,1), padding='same')(reshape1)
         B_Norm2 = layers.BatchNormalization()(DeConv1)
         Relu2 = layers.LeakyReLU()(B_Norm2)
         DeConv2 = layers.Conv2DTranspose(filters=16, kernel_size=(3,3),__
      ⇔strides=(1,1), padding='same')(Relu2)
         re = layers.Reshape((16,16,16,1))(DeConv2)
         maxi0 = layers.MaxPool3D(pool_size=(4,4,2))(re)
         B_Norm3 = layers.BatchNormalization()(maxi0)
         Relu3 = layers.LeakyReLU()(B_Norm3)
         flat0 = layers.Flatten()(Relu3)
         dense00 = layers.Dense(units=256*256, activation='tanh')(flat0)
         output_img = layers.Reshape((256, 256, 1))(dense00)
         G = Model(inputs=random_input, outputs=output_img)
         return G
[]: # Join the discriminator and generator, forming the final GAN model.
     def create_GAN():
         D = create_D()
         G = create_G()
         latent_input = keras.Input(shape=(100, ))
         img = G(latent_input)
         D.trainable = False
         score = D(img)
         GAN = Model(inputs = latent_input, outputs=score)
         GAN.compile(loss=losses.BinaryCrossentropy(),
                 optimizer=optimizers.Nadam(learning_rate=0.002), metrics=metrics.

→MeanAbsoluteError())
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Dense1 = layers.Dense(units = 16*16)(random_input)

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return D, G, GAN
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[]: | # Function returns a sample of latent points from a normal distribution.
     def getLatentSamples(num_samples) -> np.ndarray:
         latent = np.random.normal(size=100*int(num_samples))
         latent = np.reshape(a=latent, newshape=(int(num_samples), 100))
         return latent
     # Function imputs latent data into the generator and returns the fake image_
      ⇔samples outputted
     # by the generator.
     def getRealSamples(batch_size: int, train_img: np.ndarray):
         X_real, y_real = np.array(train_img.tolist() * batch_size, dtype=np.
      ofloat32).reshape((batch_size, train_img.shape[0], train_img.shape[1])), np.
      ⇔ones(shape=batch_size, dtype=np.int8)
         assert X_real.shape == (batch_size, train_img.shape[0], train_img.shape[1])
         assert y_real.shape == (batch_size, )
         return X_real, y_real
     # Function imputs latent data into the generator and returns the fake image_
      ⇔samples outputted
     # by the generator.
     def getFakeSamples(G:Model, batch_size:int):
         latent = getLatentSamples(batch size)
         assert latent.shape == (batch_size, 100)
         fake imgs = G(latent)
         fake_imgs = np.asarray(fake_imgs, dtype=np.float32)
         assert fake_imgs.shape[0] == batch_size
         fake_imgs = np.squeeze(fake_imgs)
         output_labels = np.zeros(shape=batch_size, dtype=np.int8)
         assert output_labels.shape == (batch_size, )
         return fake_imgs, output_labels
[]: # Function plots a sample of 16 'predicted' images from the trained generator.
     def pred plot(G: Model, epoch:int, train img shape: tuple[int, int]):
         shape = train_img_shape
         latent = getLatentSamples(1)
         gen_image = G(latent)
         gen_image = np.asarray(gen_image, dtype=np.float32)
         gen_image = np.squeeze(gen_image)
         assert gen_image.shape == shape
         filename = str('images/gen3-%d.png' %epoch)
         plt.imsave(filename, gen_image, cmap='gray')
         if epoch % 100 == 0:
             im=Image.open(filename)
             im.load()
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im.convert('L').show()
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[]: # Function trains the GAN (generator) model.
     def trainGAN(D:Model, G:Model, GAN:Model, num epochs: int, batchSize: int,
      strain_img: np.ndarray, train_img_shape: tuple[int, int]):
         d_loss, g_loss = [], []
         d_mae, g_mae = [], []
         pred_plot(G, 0, train_img_shape)
         for epoch in range(1, num_epochs+1):
             X_real, y_real = getRealSamples(int(batchSize/2), train_img)
             X_fake, y_fake = getFakeSamples(G, int(batchSize / 2))
             assert X_real.shape == X_fake.shape
             assert X_real.shape[0] == int(batchSize / 2)
             assert y_real.shape == y_fake.shape
             assert y_real.shape[0] == int(batchSize / 2)
             D.trainable = True
             d loss fake, mae fake = D.train on batch(X fake, y fake)
             d_loss_real, mae_real = D.train_on_batch(X_real, y_real)
             D.trainable = False
             X_latent, y_latent = getLatentSamples(batchSize), np.
      ⇔ones(shape=batchSize, dtype=np.int8)
             loss, mae = GAN.train_on_batch(X_latent, y_latent)
             d_loss_res = float((d_loss_fake + d_loss_real) / 2)
             d_mae_res = float((mae_fake + mae_real) / 2)
             d_loss.append(d_loss_res)
             d_mae.append(d_mae_res)
             g loss.append(loss)
             g_mae.append(mae)
             if (epoch \% 50 == 0) and epoch != 0:
                 pred_plot(G, epoch, train_img_shape)
                 models = (D, G, GAN)
                 filepaths = ('D_final.h5', 'G_final.h5', 'GAN_final.h5')
                 for model, name in zip(models, filepaths):
                     model.save(name)
                 print('Epoch: %d\nDiscriminator Loss: %.3f, Generator Loss: %.
      →3f\nDiscriminator MAE: %.3f, Generator MAE: %.3f' % (epoch, d_loss[-1], ⊔
      \rightarrowg_loss[-1], d_mae[-1], g_mae[-1]))
         fig, ax = plt.subplots(2,2, figsize=(11,11))
         ax[0][0].plot(list(range(len(d_loss))), d_loss, label='D_loss',u
      ⇔color='red', linestyle='dashed', linewidth=1)
         ax[0][1].plot(list(range(len(g_loss))), g_loss, label='G_loss',_
      ⇔color='red', linestyle='solid', linewidth=1)
         ax[1][0].plot(list(range(len(d_mae))), d_mae, label='D_mae', color='blue',_
      →linestyle='dashed', linewidth=1)
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ax[1][1].plot(list(range(len(g_mae))), g_mae, label='G_mae', color='blue',
linestyle='solid', linewidth=1)
fig.legend()
fig.tight_layout()
plt.savefig("loss_plots")
plt.show()
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[]: img = Image.open("orig_img_color.png")
     img.load()
     img = img.convert(mode='RGB').convert(mode='L')
     img = np.asarray(img, dtype=np.float32)
     try:
         assert img.shape == (256, 256)
     except:
         img = np.reshape(a=img, newshape=(256, 256))
     training_img = pre.minmax_scale(img.flatten(), (-1,1))
     training_img = training_img.reshape(img.shape)
     D, G, GAN = create_GAN()
     def show_summary(D, G, GAN):
         D.summary()
         G.summary()
         GAN.summary()
     verbose = input('Enter T/F for verbose output: ')
     if verbose == 'T':
         show_summary(D, G, GAN)
     trainGAN(D, G, GAN, 10000, 16, training_img, training_img.shape)
```