```
1 !mkdir ~/.kaggle
1 !cp /content/kaggle.json ~/.kaggle/
1 !kaggle datasets download -d joosthazelzet/lego-brick-images
    Warning: Your Kaggle API key is readable by other users on this system! To fix this, you can run 'chmod 600 /root/.kaggle/kaggle.json'
    Downloading lego-brick-images.zip to /content
     100% 1.00G/1.00G [00:38<00:00, 36.6MB/s]
    100% 1.00G/1.00G [00:38<00:00, 27.7MB/s]
1 !unzip /content/lego-brick-images.zip
1 # %load_ext autoreload
2 # %autoreload 2
3 import numpy as np
4 import matplotlib.pyplot as plt
6 import tensorflow as tf
7 from tensorflow.keras import (
8
      layers,
      models.
9
10
      callbacks,
11
      losses,
12
      utils,
13
      metrics,
14
      optimizers,
15 )
16
17 from drive.MyDrive.GAN_Datasets.utils import display, sample_batch
1 IMAGE_SIZE = 64
2 CHANNELS = 1
3 BATCH SIZE = 128
4 Z_DIM = 100
5 EPOCHS = 300
6 LOAD_MODEL = False
7 ADAM_BETA_1 = 0.5
8 ADAM_BETA_2 = 0.999
9 LEARNING_RATE = 0.0002
10 NOISE_PARAM = 0.1
1 # Prepare the data
2 train_data = utils.image_dataset_from_directory(
      "/content/LEGO brick images v1",
3
4
      labels=None,
      color_mode="grayscale",
5
6
      image_size=(IMAGE_SIZE, IMAGE_SIZE),
      batch_size=BATCH_SIZE,
7
8
      shuffle=True,
9
      seed=42,
10
      interpolation="bilinear",
11)
     Found 6379 files belonging to 1 classes.
1 def preprocess(img):
    img = (tf.cast(img, "float32")- 127.5)/127.5
3
    return img
5 train = train_data.map(lambda x:preprocess(x))
1 train_sample = sample_batch(train)
 2 display(train_sample)
```



Build the GAN

```
1 discriminator_input = layers.Input(shape=(IMAGE_SIZE, IMAGE_SIZE, CHANNELS))
 2 x = layers.Conv2D(64, kernel_size=4, strides=2, padding='same', use_bias=False)(discriminator_input)
3 \times = layers.LeakyReLU(0.2)(x)
 4 \times = layers.Dropout(0.3)(x)
 5 x = layers.Conv2D(128, kernel_size=4, strides=2, padding='same', use_bias=False)(x)
 6 x = layers.BatchNormalization(momentum=0.9)(x)
 7 x = layers.LeakyReLU(0.2)(x)
 8 \times = layers.Dropout(0.3)(x)
9 x = layers.Conv2D(256, kernel_size=4, strides=2, padding='same', use_bias=False)(x)
10 x = layers.BatchNormalization(momentum=0.9)(x)
11 x = layers.LeakyReLU(0.2)(x)
12 \times = layers.Dropout(0.3)(x)
13 x = layers.Conv2D(512, kernel_size=4, strides=2, padding='same', use_bias=False)(x)
14 x = layers.BatchNormalization(momentum=0.9)(x)
15 x = layers.LeakyReLU(0.2)(x)
16 \times = layers.Dropout(0.3)(x)
17 x = layers.Conv2D(1, kernel_size=4, strides=1, padding='valid', use_bias=False, activation='sigmoid',)(x)
18 discriminator_output = layers.Flatten()(x)
19 discriminator = models.Model(discriminator_input, discriminator_output)
20
21 discriminator.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 64, 64, 1)]	0
conv2d (Conv2D)	(None, 32, 32, 64)	1024
leaky_re_lu (LeakyReLU)	(None, 32, 32, 64)	0
dropout (Dropout)	(None, 32, 32, 64)	0
conv2d_1 (Conv2D)	(None, 16, 16, 128)	131072
$batch_normalization$ (BatchN $ormalization$)	(None, 16, 16, 128)	512
<pre>leaky_re_lu_1 (LeakyReLU)</pre>	(None, 16, 16, 128)	0
dropout_1 (Dropout)	(None, 16, 16, 128)	0
conv2d_2 (Conv2D)	(None, 8, 8, 256)	524288
<pre>batch_normalization_1 (Batch hormalization)</pre>	(None, 8, 8, 256)	1024
<pre>leaky_re_lu_2 (LeakyReLU)</pre>	(None, 8, 8, 256)	0
dropout_2 (Dropout)	(None, 8, 8, 256)	0
conv2d_3 (Conv2D)	(None, 4, 4, 512)	2097152
<pre>batch_normalization_2 (BatchNormalization)</pre>	(None, 4, 4, 512)	2048
<pre>leaky_re_lu_3 (LeakyReLU)</pre>	(None, 4, 4, 512)	0
dropout_3 (Dropout)	(None, 4, 4, 512)	0
conv2d_4 (Conv2D)	(None, 1, 1, 1)	8192
flatten (Flatten)	(None, 1)	0

Total params: 2,765,312 Trainable params: 2,763,520 Non-trainable params: 1,792

```
1 generator_input = layers.Input(shape=(Z_DIM,))
2 x = layers.Reshape((1,1,Z_DIM))(generator_input)
3 x = layers.Conv2DTranspose(512, kernel_size=4, strides=1, padding='valid', use_bias=False)(x)
4 \times = layers.BatchNormalization(momentum=0.9)(x)
5 x = layers.LeakyReLU(0.2)(x)
6 x = layers.Conv2DTranspose(256, kernel_size=4, strides=2, padding='same', use_bias=False)(x)
7 \times = layers.BatchNormalization(momentum=0.9)(x)
8 x = layers.LeakyReLU(0.2)(x)
9 x = layers.Conv2DTranspose(128, kernel_size=4, strides=2, padding='same', use_bias=False)(x)
10 x = layers.BatchNormalization(momentum=0.9)(x)
11 x = layers.LeakyReLU(0.2)(x)
12 x = layers.Conv2DTranspose(64, kernel_size=4, strides=2, padding='same', use_bias=False)(x)
13 x = layers.BatchNormalization(momentum=0.9)(x)
14 \times = layers.LeakyReLU(0.2)(x)
15 generator_output = layers.Conv2DTranspose(CHANNELS, kernel_size=4, strides=2, padding='same', use_bias=False, activation='tanh')(x)
16 generator = models.Model(generator_input, generator_output)
17 generator.summary()
```

Model: "model_1"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 100)]	0
reshape (Reshape)	(None, 1, 1, 100)	0
conv2d_transpose (Conv2DTranspose)	(None, 4, 4, 512)	819200
<pre>batch_normalization_3 (Batc hNormalization)</pre>	(None, 4, 4, 512)	2048
leaky_re_lu_4 (LeakyReLU)	(None, 4, 4, 512)	0
<pre>conv2d_transpose_1 (Conv2DT ranspose)</pre>	(None, 8, 8, 256)	2097152
<pre>batch_normalization_4 (BatchNormalization)</pre>	(None, 8, 8, 256)	1024
leaky_re_lu_5 (LeakyReLU)	(None, 8, 8, 256)	0
<pre>conv2d_transpose_2 (Conv2DT ranspose)</pre>	(None, 16, 16, 128)	524288
<pre>batch_normalization_5 (Batc hNormalization)</pre>	(None, 16, 16, 128)	512
leaky_re_lu_6 (LeakyReLU)	(None, 16, 16, 128)	0
<pre>conv2d_transpose_3 (Conv2DT ranspose)</pre>	(None, 32, 32, 64)	131072
<pre>batch_normalization_6 (Batc hNormalization)</pre>	(None, 32, 32, 64)	256
leaky_re_lu_7 (LeakyReLU)	(None, 32, 32, 64)	0
<pre>conv2d_transpose_4 (Conv2DT ranspose)</pre>	(None, 64, 64, 1)	1024
		=======

Total params: 3,576,576 Trainable params: 3,574,656 Non-trainable params: 1,920

```
1 class DCGAN(models.Model):
       def __init__(self, discriminator, generator, latent_dim):
           super(DCGAN, self).__init__()
 3
           self.discriminator = discriminator
 5
           self.generator = generator
           self.latent dim = latent dim
 6
 7
 8
       def compile(self, d_optimizer, g_optimizer):
 9
           super(DCGAN, self).compile()
10
           self.loss_fn = losses.BinaryCrossentropy()
           self.d_optimizer = d_optimizer
11
           self.g_optimizer = g_optimizer
12
13
           self.d_loss_metric = metrics.Mean(name="d_loss")
14
           self.d real acc metric = metrics.BinaryAccuracy(name="d real acc")
15
           self.d_fake_acc_metric = metrics.BinaryAccuracy(name="d_fake_acc")
16
           self.d_acc_metric = metrics.BinaryAccuracy(name="d_acc")
           self.g_loss_metric = metrics.Mean(name="g_loss")
17
18
           self.g_acc_metric = metrics.BinaryAccuracy(name="g_acc")
19
20
       @property
       def metrics(self):
21
22
          return [
23
              self.d_loss_metric,
24
               self.d real acc metric,
25
               self.d_fake_acc_metric,
26
               self.d_acc_metric,
27
               self.g_loss_metric,
28
               self.g_acc_metric,
29
           ]
30
      def train_step(self, real_images):
31
32
           # Sample random points in the latent space
33
           batch_size = tf.shape(real_images)[0]
           random_latent_vectors = tf.random.normal(
34
35
               shape=(batch_size, self.latent_dim)
36
37
38
           # Train the discriminator on fake images
39
           with tf.GradientTape() as gen_tape, tf.GradientTape() as disc_tape:
40
               generated_images = self.generator(
                   {\tt random\_latent\_vectors,\ training=True}
41
42
43
               real_predictions = self.discriminator(real_images, training=True)
44
               fake_predictions = self.discriminator(
45
                   generated_images, training=True
46
47
48
               real_labels = tf.ones_like(real_predictions)
               real_noisy_labels = real_labels + NOISE_PARAM * tf.random.uniform(
49
50
                   tf.shape(real_predictions)
51
52
               fake_labels = tf.zeros_like(fake_predictions)
               fake_noisy_labels = fake_labels - NOISE_PARAM * tf.random.uniform(
53
54
                   tf.shape(fake_predictions)
55
56
57
               d_real_loss = self.loss_fn(real_noisy_labels, real_predictions)
58
               d_fake_loss = self.loss_fn(fake_noisy_labels, fake_predictions)
               d_loss = (d_real_loss + d_fake_loss) / 2.0
59
60
               g_loss = self.loss_fn(real_labels, fake_predictions)
61
62
63
           gradients_of_discriminator = disc_tape.gradient(
64
               d_loss, self.discriminator.trainable_variables
65
66
           gradients_of_generator = gen_tape.gradient(
               g_loss, self.generator.trainable_variables
67
68
69
70
           self.d_optimizer.apply_gradients(
71
               zip(gradients_of_discriminator, discriminator.trainable_variables)
72
73
           self.g_optimizer.apply_gradients(
74
               \verb|zip(gradients_of_generator, generator.trainable_variables)|\\
75
76
           # Update metrics
```

```
78
           self.d_loss_metric.update_state(d_loss)
79
           self.d_real_acc_metric.update_state(real_labels, real_predictions)
           self.d_fake_acc_metric.update_state(fake_labels, fake_predictions)
80
           self.d_acc_metric.update_state(
81
               [real_labels, fake_labels], [real_predictions, fake_predictions]
82
83
84
           self.g_loss_metric.update_state(g_loss)
85
           {\tt self.g\_acc\_metric.update\_state(real\_labels, fake\_predictions)}
86
          return {m.name: m.result() for m in self.metrics}
87
1 # create a DCGAN
2 dcgan = DCGAN(discriminator=discriminator, generator=generator , latent_dim=Z_DIM)
1 if LOAD MODEL:
    dcgan.load_weights("./checkpoint/checkpoint.ckpt")
```

Train the GAN

```
1 dcgan.compile(
      d optimizer=optimizers.Adam(
3
          learning_rate=LEARNING_RATE, beta_1=ADAM_BETA_1, beta_2=ADAM_BETA_2
4
 5
      g_optimizer=optimizers.Adam(
 6
          learning_rate=LEARNING_RATE, beta_1=ADAM_BETA_1, beta_2=ADAM_BETA_2
 7
8)
1 model checkpoint callback = callbacks.ModelCheckpoint(
      filepath="./checkpoint/checkpoint.ckpt",
      save_weights_only=True,
 3
4
      save freq="epoch",
 5
      verbose=0,
6)
8 tensorboard_callback = callbacks.TensorBoard(log_dir="./logs")
10 class ImageGenerator(callbacks.Callback):
      def __init__(self, num_img, latent_dim):
11
12
          self.num_img = num_img
          self.latent_dim = latent_dim
13
14
15
      def on_epoch_end(self, epoch, logs=None):
16
          random_latent_vectors = tf.random.normal(
17
               shape=(self.num_img, self.latent_dim)
18
19
           generated_images = self.model.generator(random_latent_vectors)
20
           generated_images = generated_images * 127.5 + 127.5
           generated_images = generated_images.numpy()
21
22
           display(
23
              generated_images,
              save_to="/content/output/generated_img_%03d.png" % (epoch),
24
25
1 dcgan.fit(
2
      train,
3
      epochs=EPOCHS,
4
      callbacks=[
 5
           model_checkpoint_callback,
 6
          tensorboard_callback,
 7
           ImageGenerator(num_img=10, latent_dim=Z_DIM),
8
      ],
9)
```

```
"All channel British on Was Ballett ballett of fall
50/50 [=================] - 12s 210ms/step - d_loss: -0.7727 - d_real_ac
Epoch 286/300
Saved to /content/output/generated_img_285.png
Epoch 287/300
Saved to /content/output/generated_img_286.png
Epoch 288/300
Saved to /content/output/generated_img_287.png
 0
50/50 [=================] - 12s 211ms/step - d_loss: -0.7698 - d_real_ac
Epoch 289/300
Saved to /content/output/generated_img_288.png
Epoch 290/300
Saved to /content/output/generated_img_289.png
        50/50 [======
Epoch 291/300
Saved to /content/output/generated_img_290.png
Epoch 292/300
Saved to /content/output/generated_img_291.png
Epoch 293/300
4
```