## LandscapeGAN

January 4, 2023

```
[]: import numpy as np

[]: from tensorflow import keras

dataset = keras.utils.image_dataset_from_directory(
    "Dataset",
    label_mode=None,
    image_size=(128,128),
    batch_size=8,
    smart_resize=True
)

Found 4319 files belonging to 1 classes.

[]: dataset = dataset.map(lambda x: x/255.)

[]: import matplotlib.pyplot as plt
for x in dataset:
    plt.axis("off")
    plt.imshow((x.numpy() * 255).astype("int32")[0])
    break
```



## []: discriminator.summary() Model: "discriminator"

Layer (type) Output Shape Param #

conv2d_8 (Conv2D)	(None, 64, 64, 128)	6272
<pre>leaky_re_lu_14 (LeakyReLU)</pre>	(None, 64, 64, 128)	0
conv2d_9 (Conv2D)	(None, 32, 32, 256)	524544
<pre>leaky_re_lu_15 (LeakyReLU)</pre>	(None, 32, 32, 256)	0
conv2d_10 (Conv2D)	(None, 16, 16, 256)	1048832
<pre>leaky_re_lu_16 (LeakyReLU)</pre>	(None, 16, 16, 256)	0
flatten_2 (Flatten)	(None, 65536)	0
<pre>dropout_2 (Dropout)</pre>	(None, 65536)	0
dense_4 (Dense)	(None, 1)	65537

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Total params: 1,645,185 Trainable params: 1,645,185 Non-trainable params: 0

Layer (type)	Output Shape	 Param #
		6272
conv2d_8 (Conv2D)	(None, 64, 64, 128)	0212
leaky_re_lu_14 (LeakyReLU)	(None, 64, 64, 128)	0
conv2d_9 (Conv2D)	(None, 32, 32, 256)	524544
<pre>leaky_re_lu_15 (LeakyReLU)</pre>	(None, 32, 32, 256)	0
conv2d_10 (Conv2D)	(None, 16, 16, 256)	1048832
<pre>leaky_re_lu_16 (LeakyReLU)</pre>	(None, 16, 16, 256)	0
flatten_2 (Flatten)	(None, 65536)	0
dropout_2 (Dropout)	(None, 65536)	0
dense_4 (Dense)	(None, 1)	65537

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Total params: 1,645,185 Trainable params: 1,645,185 Non-trainable params: 0 -----

```
[]: latent_dim = 128
     generator = keras.Sequential(
            keras.Input(shape=(latent_dim,)),
            layers.Dense(8 * 8 * 128),
            layers.Reshape((8, 8, 128)),
            layers.Conv2DTranspose(128, kernel_size=4, strides=2, padding="same"),
             layers.LeakyReLU(alpha=0.2),
             layers.Conv2DTranspose(256, kernel_size=4, strides=2, padding="same"),
             layers.LeakyReLU(alpha=0.2),
             layers.Conv2DTranspose(512, kernel_size=4, strides=2, padding="same"),
             layers.LeakyReLU(alpha=0.2),
             layers.Conv2DTranspose(1024, kernel_size=4, strides=2, padding="same"),
             layers.LeakyReLU(alpha=0.2),
             layers.Conv2D(3, kernel_size=5, padding="same", activation="tanh"),
        ],
        name="generator",
```

## []: generator.summary()

Model: "generator"

Layer (type)	Output Shape	Param #
dense_5 (Dense)	(None, 8192)	1056768
reshape_2 (Reshape)	(None, 8, 8, 128)	0
<pre>conv2d_transpose_8 (Conv2DT ranspose)</pre>	(None, 16, 16, 128)	262272
<pre>leaky_re_lu_17 (LeakyReLU)</pre>	(None, 16, 16, 128)	0
<pre>conv2d_transpose_9 (Conv2DT ranspose)</pre>	(None, 32, 32, 256)	524544
leaky_re_lu_18 (LeakyReLU)	(None, 32, 32, 256)	0
conv2d_transpose_10 (Conv2D Transpose)	(None, 64, 64, 512)	2097664
leaky_re_lu_19 (LeakyReLU)	(None, 64, 64, 512)	0
conv2d_transpose_11 (Conv2D	(None, 128, 128, 1024)	8389632

## Transpose)

leaky\_re\_lu\_20 (LeakyReLU) (None, 128, 128, 1024) 0

conv2d\_11 (Conv2D) (None, 128, 128, 3) 76803

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Total params: 12,407,683 Trainable params: 12,407,683 Non-trainable params: 0

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Layer (type)	Output Shape	Param #
dense_5 (Dense)	(None, 8192)	1056768
reshape_2 (Reshape)	(None, 8, 8, 128)	0
<pre>conv2d_transpose_8 (Conv2DT ranspose)</pre>	(None, 16, 16, 128)	262272
<pre>leaky_re_lu_17 (LeakyReLU)</pre>	(None, 16, 16, 128)	0
<pre>conv2d_transpose_9 (Conv2DT ranspose)</pre>	(None, 32, 32, 256)	524544
leaky_re_lu_18 (LeakyReLU)	(None, 32, 32, 256)	0
<pre>conv2d_transpose_10 (Conv2D Transpose)</pre>	(None, 64, 64, 512)	2097664
leaky_re_lu_19 (LeakyReLU)	(None, 64, 64, 512)	0
<pre>conv2d_transpose_11 (Conv2D Transpose)</pre>	(None, 128, 128, 1024)	8389632
<pre>leaky_re_lu_20 (LeakyReLU)</pre>	(None, 128, 128, 1024)	0
conv2d_11 (Conv2D)	(None, 128, 128, 3)	76803

Total params: 12,407,683 Trainable params: 12,407,683 Non-trainable params: 0

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```
[]: import tensorflow as tf
     class GAN(keras.Model):
         def __init__(self, discriminator, generator, latent_dim):
             super().__init__()
             self.discriminator = discriminator
             self.generator = generator
             self.latent_dim = latent_dim
             self.d_loss_metric = keras.metrics.Mean(name="d_loss")
             self.g_loss_metric = keras.metrics.Mean(name="g_loss")
         def compile(self, d_optimizer, g_optimizer, loss_fn):
             super(GAN, self).compile()
             self.d_optimizer = d_optimizer
             self.g_optimizer = g_optimizer
             self.loss_fn = loss_fn
         @property
         def metrics(self):
             return [self.d_loss_metric, self.g_loss_metric]
         def train_step(self, real_images):
             batch_size = tf.shape(real_images)[0]
             random_latent_vectors = tf.random.normal(
                 shape=(batch size, self.latent dim))
             generated_images = self.generator(random_latent_vectors)
             combined images = tf.concat([generated images, real images], axis=0)
             labels = tf.concat(
                 [tf.ones((batch_size, 1)), tf.zeros((batch_size, 1))],
                 axis=0
             labels += 0.1 * tf.random.uniform(tf.shape(labels))
             with tf.GradientTape() as tape:
                 predictions = self.discriminator(combined_images)
                 d_loss = self.loss_fn(labels, predictions)
             grads = tape.gradient(d_loss, self.discriminator.trainable_weights)
             self.d_optimizer.apply_gradients(
                 zip(grads, self.discriminator.trainable_weights)
             )
             random_latent_vectors = tf.random.normal(
                 shape=(batch_size, self.latent_dim))
             misleading_labels = tf.zeros((batch_size, 1))
             with tf.GradientTape() as tape:
                 predictions = self.discriminator(
```

```
self.generator(random_latent_vectors))
               g_loss = self.loss_fn(misleading_labels, predictions)
           grads = tape.gradient(g_loss, self.generator.trainable_weights)
           self.g_optimizer.apply_gradients(
               zip(grads, self.generator.trainable_weights))
           self.d_loss_metric.update_state(d_loss)
           self.g_loss_metric.update_state(g_loss)
           return {"d_loss": self.d_loss_metric.result(),
                   "g_loss": self.g_loss_metric.result()}
[]: class GANMonitor(keras.callbacks.Callback):
        def __init__(self, num_img=3, latent_dim=128):
           self.num_img = num_img
           self.latent_dim = latent_dim
        def on_epoch_end(self, epoch, logs=None):
           random_latent_vectors = tf.random.normal(shape=(self.num_img, self.
     →latent_dim))
           generated_images = self.model.generator(random_latent_vectors)
           generated_images *= 255
           generated_images.numpy()
           for i in range(self.num_img):
               img = keras.utils.array_to_img(generated_images[i])
               img.save(f"Generated Images\generated_img_{epoch:03d}_{i}.png")
[]: epochs = 50
    gan = GAN(discriminator=discriminator, generator=generator,_
     →latent_dim=latent_dim)
    gan.compile(
        d_optimizer=keras.optimizers.Adam(learning_rate=0.000001),
        g_optimizer=keras.optimizers.Adam(learning_rate=0.00001),
        loss_fn=keras.losses.BinaryCrossentropy(),
    )
    gan.fit(
        dataset, epochs=epochs, callbacks=[GANMonitor(num_img=10,_
     ⇔latent_dim=latent_dim)]
    )
   Epoch 1/50
   g_loss: 0.6741
   Epoch 2/50
   g_loss: 0.6680
```

Epoch 3/50

```
g_loss: 0.7112
Epoch 4/50
g loss: 0.6676
Epoch 5/50
g_loss: 0.6794
Epoch 6/50
g_loss: 0.6495
Epoch 7/50
g_loss: 0.6665
Epoch 8/50
g_loss: 0.6772
Epoch 9/50
g loss: 0.6957
Epoch 10/50
g_loss: 0.6954
Epoch 11/50
g_loss: 0.6967
Epoch 12/50
g_loss: 0.7026
Epoch 13/50
g_loss: 0.7039
Epoch 14/50
g loss: 0.7097
Epoch 15/50
g_loss: 0.7079
Epoch 16/50
g_loss: 0.7085
Epoch 17/50
g_loss: 0.7081
Epoch 18/50
g_loss: 0.7088
Epoch 19/50
```

```
g_loss: 0.7102
Epoch 20/50
g loss: 0.7177
Epoch 21/50
g_loss: 0.7127
Epoch 22/50
g_loss: 0.7150
Epoch 23/50
g_loss: 0.7162
Epoch 24/50
g_loss: 0.7187
Epoch 25/50
g loss: 0.7272
Epoch 26/50
g_loss: 0.7213
Epoch 27/50
g_loss: 0.7169
Epoch 28/50
g_loss: 0.7250
Epoch 29/50
g_loss: 0.7213
Epoch 30/50
g loss: 0.7225
Epoch 31/50
g_loss: 0.7292
Epoch 32/50
g_loss: 0.7300
Epoch 33/50
g_loss: 0.7254
Epoch 34/50
g_loss: 0.7343
Epoch 35/50
```

```
g_loss: 0.7276
Epoch 36/50
g loss: 0.7263
Epoch 37/50
g_loss: 0.7383
Epoch 38/50
g_loss: 0.7312
Epoch 39/50
g_loss: 0.7273
Epoch 40/50
g_loss: 0.7387
Epoch 41/50
g loss: 0.7282
Epoch 42/50
g_loss: 0.7290
Epoch 43/50
g_loss: 0.7322
Epoch 44/50
g_loss: 0.7353
Epoch 45/50
g_loss: 0.7364
Epoch 46/50
g loss: 0.7319
Epoch 47/50
g_loss: 0.7366
Epoch 48/50
g_loss: 0.7372
Epoch 49/50
g_loss: 0.7343
Epoch 50/50
g_loss: 0.7275
```

```
[]: <keras.callbacks.History at 0x176b83c8940>
```

```
[]: generator.save('savedmodels\GeneratorTanh04/01')
```

WARNING:tensorflow:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile\_metrics` will be empty until you train or evaluate the model.

WARNING:absl:Found untraced functions such as \_jit\_compiled\_convolution\_op, \_jit\_compiled\_convolution\_op, \_jit\_compiled\_convolution\_op, \_jit\_compiled\_convolution\_op while saving (showing 5 of 5). These functions will not be directly callable after loading.

INFO:tensorflow:Assets written to: savedmodels\GeneratorTanh04/01\assets

INFO:tensorflow:Assets written to: savedmodels\GeneratorTanh04/01\assets

```
[]: import tensorflow as tf
new_model = tf.keras.models.load_model('savedmodels\GeneratorTanh04/01')
```

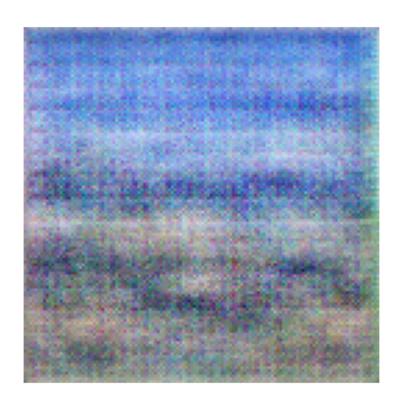
WARNING:tensorflow:No training configuration found in save file, so the model was \*not\* compiled. Compile it manually.

WARNING:tensorflow:No training configuration found in save file, so the model was \*not\* compiled. Compile it manually.

```
[]: num = 1
    random_vector = tf.random.normal(shape=(num, 256))
    images = new_model(random_vector)
    images *= 255
    tf.shape(images[0])
```

[]: <tf.Tensor: shape=(3,), dtype=int32, numpy=array([256, 256, 3])>

```
[]: for x in range(num):
    img = keras.utils.array_to_img(images[x])
    plt.axis("off")
    plt.imshow(img)
```



```
[]: new_model.summary()
[]: import tensorflow as tf
     from tensorflow import keras
     from tensorflow.keras import layers
     from tensorflow.keras import mixed_precision
     policy = mixed_precision.Policy('float32')
    mixed_precision.set_global_policy(policy)
     mixed_precision.set_global_policy('float32')
     print('Compute dtype: %s' % policy.compute_dtype)
    print('Variable dtype: %s' % policy.variable_dtype)
    Compute dtype: float32
    Variable dtype: float32
[ ]: num = 1
     random_vector = tf.random.normal(shape=(num, 128))
     images = new_model(random_vector)
     images *= 255
     print(tf.shape(images[0]))
```

```
for x in range(num):
    img = keras.utils.array_to_img(images[x])
    plt.axis("off")
    plt.imshow(img)
```

tf.Tensor([128 128 3], shape=(3,), dtype=int32)

