Image_Colorization_Generic_GAN_Model

March 24, 2023

1 Image Colorization with GANs

```
[]: from PIL import Image
from sklearn.model_selection import train_test_split
import tensorflow as tf
import numpy as np
from matplotlib import image
from matplotlib import pyplot as plt
import os
from tensorflow import keras
```

1.1 Data load

```
[]: batch size = 64
     img_size = 120
     dataset_split = 2500
     master_dir = 'Endoscopy/images/'
     x = []
     y = []
     for image_file in os.listdir( master_dir )[ 0 : dataset_split ]:
         rgb_image = Image.open( os.path.join( master_dir , image_file ) ).resize( (__
      →img_size , img_size ) )
         rgb_img_array = (np.asarray( rgb_image ) ) / 255
         gray_image = rgb_image.convert( 'L' )
        gray_img_array = ( np.asarray( gray_image ).reshape( ( img_size , img_size_
      →, 1 ) ) / 255
         x.append( gray_img_array )
         y.append( rgb_img_array )
     # Train-test splitting
     train_x, test_x, train_y, test_y = train_test_split( np.array(x) , np.array(y)_
     →, test_size=0.1 )
     # Construct tf.data.Dataset object
     dataset = tf.data.Dataset.from_tensor_slices( ( train x , train y ) )
     dataset = dataset.batch( batch_size )
```

1.1.1 GAN: Generator

```
[3]: def get_generator_model():
         inputs = tf.keras.layers.Input( shape=( img_size , img_size , 1 ) )
         conv1 = tf.keras.layers.Conv2D( 16 , kernel_size=( 5 , 5 ) , strides=1 )(__
      ⇒inputs )
         conv1 = tf.keras.layers.LeakyReLU()( conv1 )
         conv1 = tf.keras.layers.Conv2D( 32 , kernel_size=( 3 , 3 ) , strides=1)(__
      ⇔conv1 )
         conv1 = tf.keras.layers.LeakyReLU()( conv1 )
         conv1 = tf.keras.layers.Conv2D( 32 , kernel_size=( 3 , 3 ) , strides=1)(__
         conv1 = tf.keras.layers.LeakyReLU()( conv1 )
         conv2 = tf.keras.layers.Conv2D( 32 , kernel_size=( 5 , 5 ) , strides=1)(__
      ⇔conv1 )
         conv2 = tf.keras.layers.LeakyReLU()( conv2 )
         conv2 = tf.keras.layers.Conv2D(64, kernel_size=(3,3), strides=1)(_{\square}
      ⇔conv2 )
         conv2 = tf.keras.layers.LeakyReLU()( conv2 )
         conv2 = tf.keras.layers.Conv2D( 64 , kernel_size=( 3 , 3 ) , strides=1 )(__
      ⇔conv2 )
         conv2 = tf.keras.layers.LeakyReLU()( conv2 )
         conv3 = tf.keras.layers.Conv2D( 64 , kernel_size=( 5 , 5 ) , strides=1 )(__
      ⇔conv2 )
         conv3 = tf.keras.layers.LeakyReLU()( conv3 )
         conv3 = tf.keras.layers.Conv2D( 128 , kernel_size=( 3 , 3 ) , strides=1 )(__
      ⇔conv3 )
         conv3 = tf.keras.layers.LeakyReLU()( conv3 )
         conv3 = tf.keras.layers.Conv2D( 128 , kernel_size=( 3 , 3 ) , strides=1 )(__
      ⇔conv3 )
         conv3 = tf.keras.layers.LeakyReLU()( conv3 )
         bottleneck = tf.keras.layers.Conv2D( 128 , kernel_size=( 3 , 3 ) ,
      ⇔strides=1 , activation='tanh' , padding='same' )( conv3 )
         concat_1 = tf.keras.layers.Concatenate()( [ bottleneck , conv3 ] )
         conv_up_3 = tf.keras.layers.Conv2DTranspose( 128 , kernel_size=( 3 , 3 ) ,__
      ⇔strides=1 , activation='relu' )( concat_1 )
         conv_up_3 = tf.keras.layers.Conv2DTranspose( 128 , kernel_size=( 3 , 3 ) ,__
      ⇔strides=1 , activation='relu' )( conv_up_3 )
         conv_up_3 = tf.keras.layers.Conv2DTranspose(64 , kernel_size=(5 , 5 ) ,_
      ⇔strides=1 , activation='relu' )( conv_up_3 )
```

```
concat_2 = tf.keras.layers.Concatenate()( [ conv_up_3 , conv2 ] )
  conv_up_2 = tf.keras.layers.Conv2DTranspose(64 , kernel_size=(3 , 3) ,
⇔strides=1 , activation='relu' )( concat_2 )
  conv_up_2 = tf.keras.layers.Conv2DTranspose(64 , kernel_size=(3 , 3 ) ,
⇔strides=1 , activation='relu' )( conv_up_2 )
  conv_up_2 = tf.keras.layers.Conv2DTranspose( 32 , kernel_size=( 5 , 5 ) ,__
⇔strides=1 , activation='relu' )( conv_up_2 )
  concat_3 = tf.keras.layers.Concatenate()([conv_up_2, conv1])
  conv_up_1 = tf.keras.layers.Conv2DTranspose( 32 , kernel_size=( 3 , 3 ) ,_
⇔strides=1 , activation='relu')( concat_3 )
  conv_up_1 = tf.keras.layers.Conv2DTranspose( 32 , kernel_size=( 3 , 3 ) ,
strides=1 , activation='relu')( conv_up_1 )
  conv_up_1 = tf.keras.layers.Conv2DTranspose(3, kernel_size=(5,5),_
strides=1 , activation='relu')( conv_up_1 )
  model = tf.keras.models.Model( inputs , conv_up_1 )
  return model
```

1.1.2 GAN:Discriminator

```
[4]: def get_discriminator_model():
        layers = [
             tf.keras.layers.Conv2D( 32 , kernel_size=( 7 , 7 ) , strides=1 ,__
      →activation='relu' , input_shape=( 120 , 120 , 3 ) ),
             tf.keras.layers.Conv2D( 32 , kernel_size=( 7, 7 ) , strides=1,__
      ⇔activation='relu' ),
            tf.keras.layers.MaxPooling2D(),
             tf.keras.layers.Conv2D( 64 , kernel_size=( 5 , 5 ) , strides=1,__
      ⇒activation='relu' ),
             tf.keras.layers.Conv2D( 64 , kernel_size=( 5 , 5 ) , strides=1,__
      ⇔activation='relu' ),
             tf.keras.layers.MaxPooling2D(),
             tf.keras.layers.Conv2D( 128 , kernel size=( 3 , 3 ) , strides=1,
      ⇔activation='relu' ),
             tf.keras.layers.Conv2D( 128 , kernel_size=( 3 , 3 ) , strides=1,__
      ⇔activation='relu' ),
             tf.keras.layers.MaxPooling2D(),
             tf.keras.layers.Conv2D( 256 , kernel_size=( 3 , 3 ) , strides=1,__
      ⇔activation='relu' ),
             tf.keras.layers.Conv2D( 256 , kernel_size=( 3 , 3 ) , strides=1,__
      ⇒activation='relu' ),
            tf.keras.layers.MaxPooling2D(),
             tf.keras.layers.Flatten(),
            tf.keras.layers.Dense(512, activation='relu'),
             tf.keras.layers.Dense( 128 , activation='relu' ) ,
```

```
tf.keras.layers.Dense( 16 , activation='relu' ) ,
   tf.keras.layers.Dense( 1 , activation='sigmoid' )
]
model = tf.keras.models.Sequential( layers )
return model
```

1.1.3 Loss Functions

```
[5]: cross_entropy = tf.keras.losses.BinaryCrossentropy()
     mse = tf.keras.losses.MeanSquaredError()
     def discriminator_loss(real_output, fake_output):
         real_loss = cross_entropy(tf.ones_like(real_output) - tf.random.uniform(__
      shape=real_output.shape , maxval=0.1 ) , real_output)
         fake_loss = cross_entropy(tf.zeros_like(fake_output) + tf.random.uniform(__
      ⇒shape=fake_output.shape , maxval=0.1 ) , fake_output)
         total_loss = real_loss + fake_loss
         return total_loss
     def generator_loss(fake_output , real_y):
         real_y = tf.cast( real_y , 'float32' )
         return mse( fake_output , real_y )
     generator optimizer = tf.keras.optimizers.Adam( 0.0005 )
     discriminator_optimizer = tf.keras.optimizers.Adam( 0.0005 )
     generator = get_generator_model()
     discriminator = get_discriminator_model()
```

1.2 Training GAN

```
[6]: @tf.function
    def train_step( input_x , real_y ):

        with tf.GradientTape() as gen_tape, tf.GradientTape() as disc_tape:
            generated_images = generator( input_x , training=True)
            real_output = discriminator( real_y, training=True)
            generated_output = discriminator(generated_images, training=True)

# L2 Loss
            gen_loss = generator_loss( generated_images , real_y )
            disc_loss = discriminator_loss( real_output, generated_output )

# Compute the gradients
            gradients_of_generator = gen_tape.gradient(gen_loss, generator.

-trainable_variables)
```

1.3 Results

plt.show()

```
[17]: y = generator( test_x[0 : ] ).numpy()

[18]: H,W=[120,120]
    for i in range(len(test_x)):
        plt.figure(figsize=(10,10))
        or_image = plt.subplot(3,3,1)
        or_image.set_title('Grayscale Input', fontsize=16)
        plt.imshow( test_x[i].reshape((H,W)) , cmap='gray' )

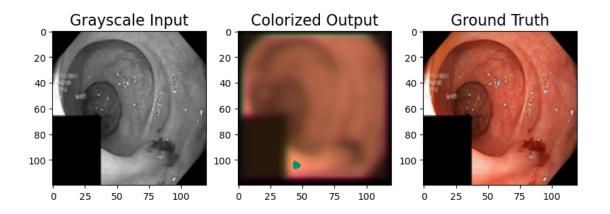
        in_image = plt.subplot(3,3,2)
```

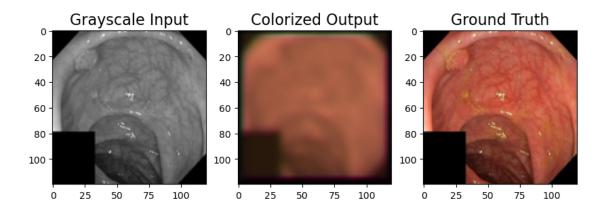
```
image = Image.fromarray( ( y[i] * 255 ).astype( 'uint8' ) ).resize( ( H , W_

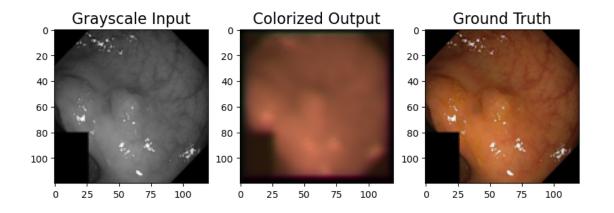
image = np.asarray( image )
   in_image.set_title('Colorized Output', fontsize=16)
   plt.imshow( image )

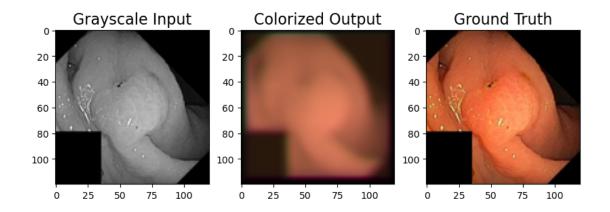
ou_image = plt.subplot(3,3,3)
   image = Image.fromarray( ( test_y[i] * 255 ).astype( 'uint8' ) ).resize( (_U

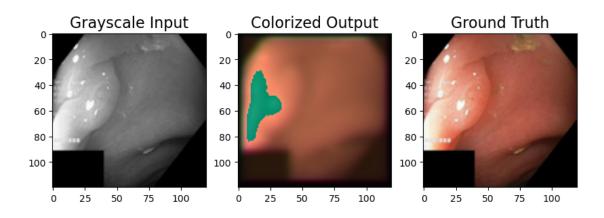
H , W ) )
   ou_image.set_title('Ground Truth', fontsize=16)
   plt.imshow( image )
```

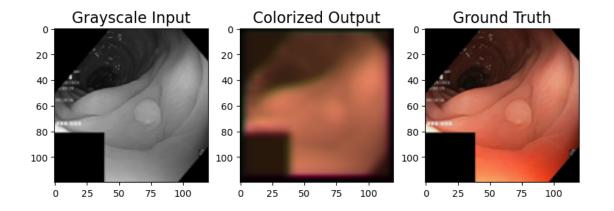


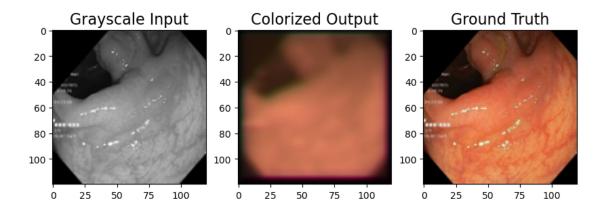


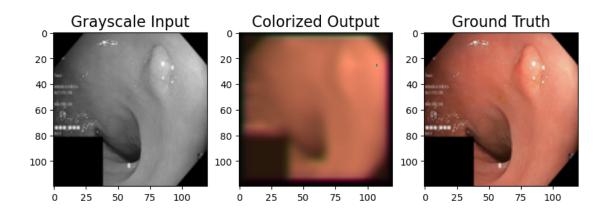


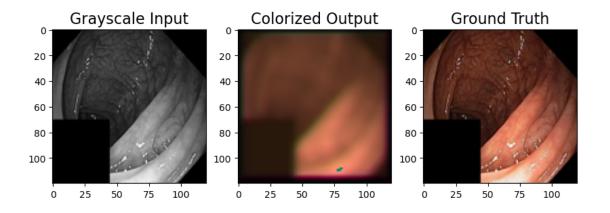


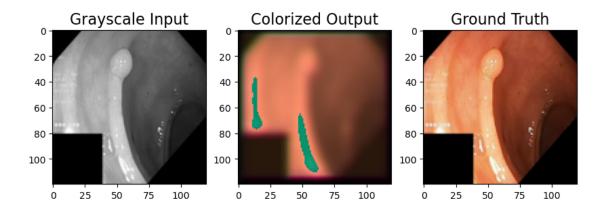


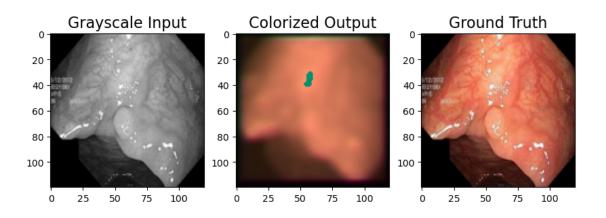


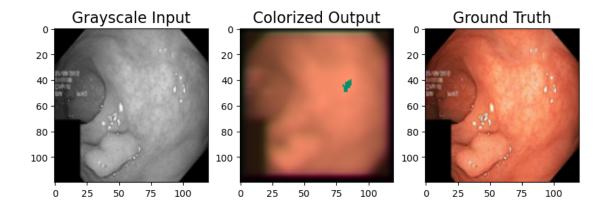


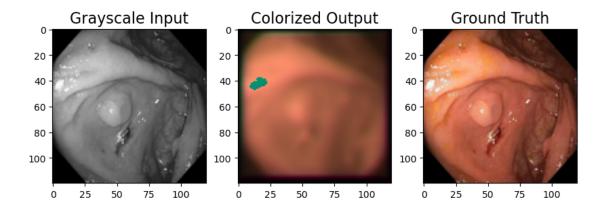


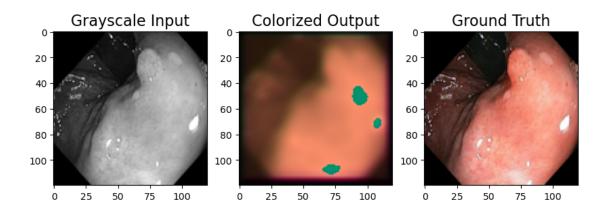


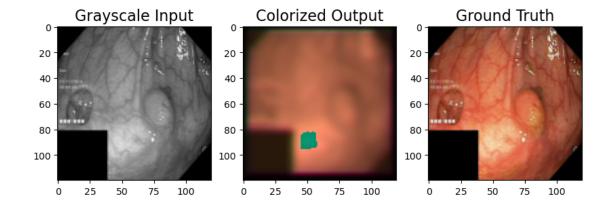


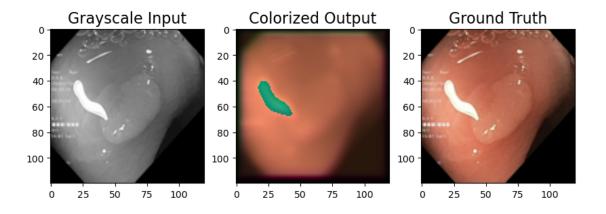


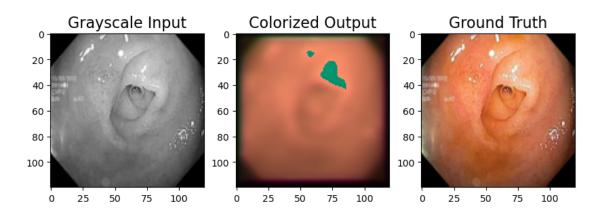


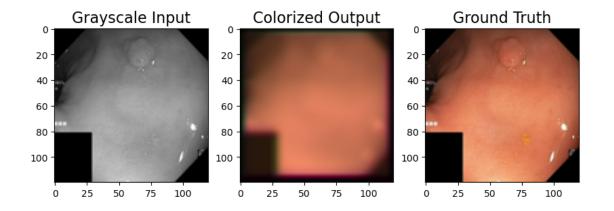


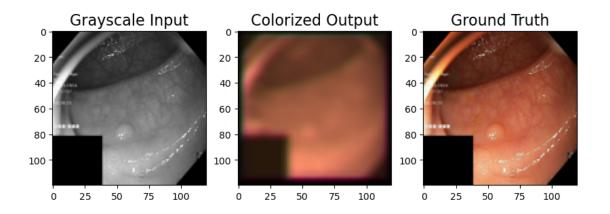


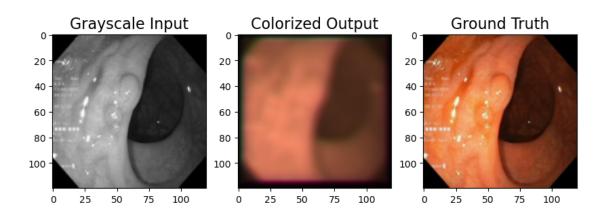












[]: