```
!rm -rf clone && git clone https://github.com/Bileth/Grayscale-image-colorization clone &&
     Cloning into 'clone'...
     remote: Enumerating objects: 14317, done.
     remote: Counting objects: 100% (14317/14317), done.
     remote: Compressing objects: 100% (14311/14311), done.
     remote: Total 14317 (delta 5), reused 14304 (delta 1), pack-reused 0
     Receiving objects: 100% (14317/14317), 230.02 MiB | 21.16 MiB/s, done.
     Resolving deltas: 100% (5/5), done.
     Checking out files: 100% (22296/22296), done.
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
# Veličina serije koju ćemo koristiti za obuku
batch_size = 64
# Size of the image required to train our model
img size = 120
# Veličina slika potrebnih za treniranje našeg modela
dataset split = 2500
master dir = 'train/images/color/'
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 ,
    # Normalize the RGB image array
    rgb img array = (np.asarray( rgb image ) ) / 255
    gray image = rgb image.convert( 'L' )
    # Normaliziranje RGB niza slika
    gray img array = ( np.asarray( gray image ).reshape( ( img size , img size , 1 ) ) ) /
    # Nadodavnje oba niza slika
    x.append( gray img array )
    y.append( rgb_img_array )
# Train-test podjela
train_x, test_x, train_y, test_y = train_test_split( np.array(x) , np.array(y) , test_size
# Konstrukcija tf.data.Dataset object
dataset = tf.data.Dataset.from_tensor_slices( ( train_x , train_y ) )
dataset = dataset.batch( batch size )
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 )
 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 )( 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 , activatio
 concat 1 = tf.keras.layers.Concatenate()( [ bottleneck , conv3 ] )
 conv_up_3 = tf.keras.layers.Conv2DTranspose( 128 , kernel_size=( 3 , 3 ) , strides=1 , a
 conv_up_3 = tf.keras.layers.Conv2DTranspose( 128 , kernel_size=( 3 , 3 ) , strides=1 , a
 conv_up_3 = tf.keras.layers.Conv2DTranspose( 64 , kernel_size=( 5 , 5 ) , strides=1 , ac
 concat_2 = tf.keras.layers.Concatenate()( [ conv_up_3 , conv2 ] )
 conv_up_2 = tf.keras.layers.Conv2DTranspose( 64 , kernel_size=( 3 , 3 ) , strides=1 , ac
 conv_up_2 = tf.keras.layers.Conv2DTranspose( 64 , kernel_size=( 3 , 3 ) , strides=1 , ac
 conv_up_2 = tf.keras.layers.Conv2DTranspose( 32 , kernel_size=( 5 , 5 ) , strides=1 , ac
 concat_3 = tf.keras.layers.Concatenate()( [ conv_up_2 , conv1 ] )
 conv up 1 = tf.keras.layers.Conv2DTranspose( 32 , kernel size=( 3 , 3 ) , strides=1 , ac
 conv_up_1 = tf.keras.layers.Conv2DTranspose( 32 , kernel_size=( 3 , 3 ) , strides=1 , ac
 conv up 1 = tf.keras.layers.Conv2DTranspose( 3 , kernel size=( 5 , 5 ) , strides=1 , act
 model = tf.keras.models.Model( inputs , conv up 1 )
 return model
def get_discriminator_model():
 lavers = [
           tf.keras.layers.Conv2D( 32 , kernel_size=( 7 , 7 ) , strides=1 , activation='r
           tf.keras.layers.Conv2D( 32 , kernel_size=( 7, 7 ) , strides=1, activation='rel
           tf.keras.layers.MaxPooling2D(),
           tf.keras.layers.Conv2D( 64 , kernel_size=( 5 , 5 ) , strides=1, activation='re
           tf.keras.layers.Conv2D( 64 , kernel_size=( 5 , 5 ) , strides=1, activation='re
           tf.keras.layers.MaxPooling2D(),
           tf.keras.layers.Conv2D( 128 , kernel_size=( 3 , 3 ) , strides=1, activation='r
           tf.keras.layers.Conv2D( 128 , kernel_size=( 3 , 3 ) , strides=1, activation='r
           tf.keras.layers.MaxPooling2D(),
```

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tf.keras.layers.Conv2D( 256 , kernel_size=( 3 , 3 ) , strides=1, activation='r
            tf.keras.layers.Conv2D( 256 , kernel size=( 3 , 3 ) , strides=1, activation='r
            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
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_ou
    fake loss = cross entropy(tf.zeros like(fake output) + tf.random.uniform( shape=fake o
    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()
@tf.function
def train_step( input_x , real_y ):
    with tf.GradientTape() as gen tape, tf.GradientTape() as disc tape:
        # generiranje slike -> G( x )
        generated_images = generator( input_x , training=True)
        # Vjerojatnost da je slika prava (točna) -> D( x )
        real_output = discriminator( real_y, training=True)
        # Vjerojatnost da je slika generirana -> D( G( x ) )
        generated_output = discriminator(generated_images, training=True)
        # L2 gubitak -> || y - G(x) ||^2
        gen_loss = generator_loss( generated_images , real_y )
        # Log gubitka za diskriminator
        disc_loss = discriminator_loss( real_output, generated_output )
    #tf.keras.backend.print tensor( tf.keras.backend.mean( gen loss ) )
    #tf.keras.backend.print tensor( gen loss + disc loss )
    # Računanje gradijenata
    gradients_of_generator = gen_tape.gradient(gen_loss, generator.trainable_variables)
    gradients_of_discriminator = disc_tape.gradient(disc_loss, discriminator.trainable_var
```

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(64, 120, 120, 1)

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optimizacija sa Adam-om

```
generator optimizer.apply gradients(zip(gradients of generator, generator.trainable va
    discriminator optimizer.apply gradients(zip(gradients of discriminator, discriminator.
num_epochs = 150
for e in range( num_epochs ):
    print( e )
    for ( x , y ) in dataset:
        # Ovjde ( x , y ) predstavlja skup iz našeg skupa podataka za obuku.
        print( x.shape )
        train_step( x , y )
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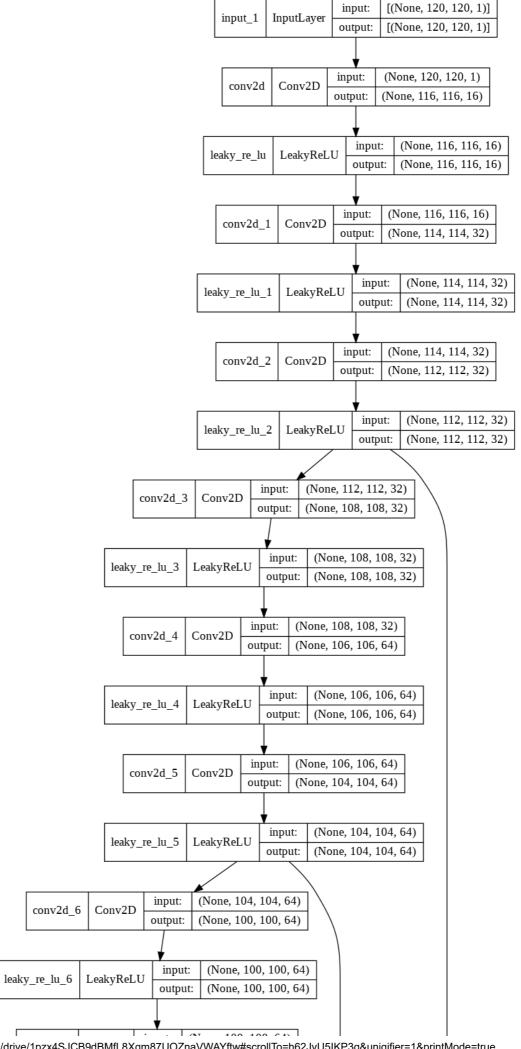
(64, 120, 120, 1)

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(64, 120, 120, 1)
```

```
generator.compile(optimizer='adam', loss='binary_crossentropy')
tf.keras.utils.plot_model(generator, 'encoder_compress.png', show_shapes=True)
```



```
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( ( 1024 , 1024 )
ou_image.set_title('Ground Truth', fontsize=16)
plt.imshow( image )

plt.savefig("plots.pdf")
plt.show()
```