Sprawozdanie Lab 11

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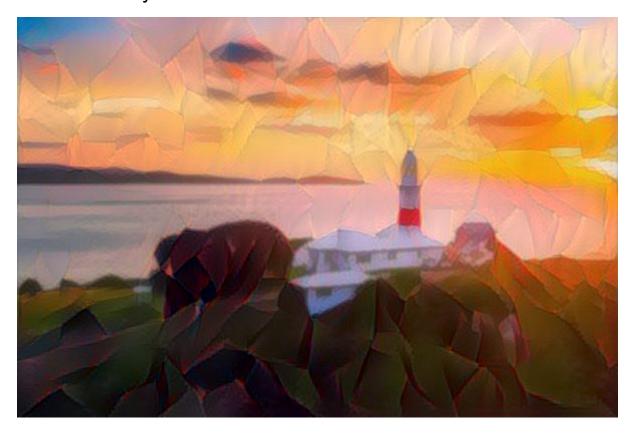
Content:



Style:



Efekt końcowy:



Kod:

https://colab.research.google.com/github/keras-team/keras-io/blob/master/examples/generative/ipynb/neural_style_transfer.ipynb#scrollTo=le3Xjou7nITA

Rozwinięcie:

```
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.applications import vgg19

base_image_path = keras.utils.get_file("paris.jpg",
"https://editorial.pxcrush.net/caravancampingsales/general/editorial/lighthouse-4.jpg?width=375")
style reference image path = keras.utils.get file(
```

```
"starry night.jpg",
"https://i.pinimg.com/originals/cf/d3/2e/cfd32ecb5df69b18dad171345dd8c1
8a.jpg"
result prefix = "paris generated"
# Weights of the different loss components
total variation weight = 1e-6
style weight = 1e-6
content_weight = 2.5e-8
# Dimensions of the generated picture.
width, height =
\verb|keras.preprocessing.image.load_img(base_image_path).size|
img nrows = 400
img_ncols = int(width * img_nrows / height)
from IPython.display import Image, display
display(Image(base image path))
display(Image(style reference image path))
def preprocess_image(image_path):
    # Util function to open, resize and format pictures into
appropriate tensors
    img = keras.preprocessing.image.load_img(
        image path, target size=(img nrows, img ncols)
    img = keras.preprocessing.image.img to array(img)
    img = np.expand_dims(img, axis=0)
    img = vgg19.preprocess input(img)
```

```
def deprocess image(x):
    # Util function to convert a tensor into a valid image
    x = x.reshape((img nrows, img ncols, 3))
    # Remove zero-center by mean pixel
    x[:, :, 0] += 103.939
    x[:, :, 1] += 116.779
    x[:, :, 2] += 123.68
    # 'BGR'->'RGB'
    x = x[:, :, ::-1]
    x = np.clip(x, 0, 255).astype("uint8")
    return x
# The gram matrix of an image tensor (feature-wise outer product)
def gram matrix(x):
    x = tf.transpose(x, (2, 0, 1))
    features = tf.reshape(x, (tf.shape(x)[0], -1))
    gram = tf.matmul(features, tf.transpose(features))
    return gram
# The "style loss" is designed to maintain
# the style of the reference image in the generated image.
# It is based on the gram matrices (which capture style) of
# feature maps from the style reference image
# and from the generated image
def style loss(style, combination):
    S = gram matrix(style)
    C = gram matrix(combination)
    channels = 3
    size = img_nrows * img_ncols
```

return tf.convert to tensor(img)

```
return tf.reduce sum(tf.square(S - C)) / (4.0 * (channels ** 2) *
(size ** 2))
# An auxiliary loss function
# designed to maintain the "content" of the
# base image in the generated image
def content loss(base, combination):
   return tf.reduce sum(tf.square(combination - base))
# The 3rd loss function, total variation loss,
# designed to keep the generated image locally coherent
def total variation loss(x):
   a = tf.square(
       x[:, : img nrows - 1, : img_ncols - 1, :] - x[:, 1:, :]
img ncols - 1, :]
   b = tf.square(
       x[:, : img nrows - 1, : img ncols - 1, :] - x[:, : img nrows -
1, 1:, :]
   return tf.reduce sum(tf.pow(a + b, 1.25))
# Build a VGG19 model loaded with pre-trained ImageNet weights
model = vgg19.VGG19(weights="imagenet", include_top=False)
# Get the symbolic outputs of each "key" layer (we gave them unique
names).
outputs_dict = dict([(layer.name, layer.output) for layer in
model.layers])
# Set up a model that returns the activation values for every layer in
# VGG19 (as a dict).
```

```
feature extractor = keras.Model(inputs=model.inputs,
outputs=outputs dict)
# List of layers to use for the style loss.
style layer names = [
    "block1 conv1",
    "block2 conv1",
    "block3 conv1",
    "block4 conv1",
    "block5 conv1",
# The layer to use for the content loss.
content layer name = "block5 conv2"
def compute loss (combination image, base image, style reference image):
    input tensor = tf.concat(
        [base image, style reference image, combination image], axis=0
    features = feature extractor(input tensor)
    # Initialize the loss
    loss = tf.zeros(shape=())
    # Add content loss
    layer features = features[content layer name]
    base image features = layer features[0, :, :, :]
    combination features = layer features[2, :, :, :]
    loss = loss + content weight * content loss(
        base image features, combination features
    # Add style loss
    for layer_name in style_layer_names:
        layer features = features[layer name]
        style reference features = layer features[1, :, :, :]
        combination features = layer features[2, :, :, :]
        sl = style loss(style reference features, combination features)
        loss += (style weight / len(style layer names)) * sl
```

```
# Add total variation loss
    loss += total variation weight *
total variation loss(combination image)
    return loss
@tf.function
def compute loss and grads (combination image, base image,
style reference image):
    with tf.GradientTape() as tape:
        loss = compute loss(combination image, base image,
style reference image)
    grads = tape.gradient(loss, combination image)
    return loss, grads
optimizer = keras.optimizers.SGD(
    keras.optimizers.schedules.ExponentialDecay(
        initial_learning_rate=100.0, decay_steps=100, decay_rate=0.96
   )
)
base_image = preprocess_image(base_image_path)
style reference image = preprocess image(style reference image path)
combination image = tf.Variable(preprocess image(base image path))
iterations = 4000
for i in range(1, iterations + 1):
    loss, grads = compute loss and grads(
        combination image, base image, style reference image
    optimizer.apply gradients([(grads, combination image)])
    if i % 100 == 0:
        print("Iteration %d: loss=%.2f" % (i, loss))
```

```
img = deprocess_image(combination_image.numpy())
    fname = result_prefix + "_at_iteration_%d.png" % i
    keras.preprocessing.image.save_img(fname, img)

display(Image(result_prefix + "_at_iteration_4000.png"))
```