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### Part of this code is due to the MatConvNet team and is used to load the parameters of the pr

import os
import sys
import scipy.io
import scipy.misc
import matplotlib.pyplot as plt
from matplotlib.pyplot import imshow
from PIL import Image
from nst_utils import *

import numpy as np
import tensorflow as tf

class CONFIG:
    IMAGE_WIDTH = 400
    IMAGE_HEIGHT = 300
    COLOR_CHANNELS = 3
    NOISE_RATIO = 0.6
    MEANS = np.array([123.68, 116.779, 103.939]).reshape((1,1,1,3))
    VGG_MODEL = 'pretrained-model/imagenet-vgg-verydeep-19.mat' # Pick the VGG 19-layer model by
    STYLE_IMAGE = 'images/stone_style.jpg' # Style image to use.
    CONTENT_IMAGE = 'images/content300.jpg' # Content image to use.
    OUTPUT_DIR = 'output/'

def load_vgg_model(path):
    """
    Returns a model for the purpose of 'painting' the picture.
    Takes only the convolution layer weights and wrap using the TensorFlow
    Conv2d, Relu and AveragePooling layer. VGG actually uses maxpool but
    the paper indicates that using AveragePooling yields better results.
    The last few fully connected layers are not used.
    Here is the detailed configuration of the VGG model:
        0 is conv1_1 (3, 3, 3, 64)
        1 is relu
        2 is conv1_2 (3, 3, 64, 64)
        3 is relu
        4 is maxpool
        5 is conv2_1 (3, 3, 64, 128)
        6 is relu
        7 is conv2_2 (3, 3, 128, 128)
        8 is relu
        9 is maxpool
        10 is conv3_1 (3, 3, 128, 256)
        11 is relu
        12 is conv3_2 (3, 3, 256, 256)
        13 is relu
        14 is conv3_3 (3, 3, 256, 256)
        15 is relu
        16 is conv3_4 (3, 3, 256, 256)
        17 is relu
        18 is maxpool
        19 is conv4_1 (3, 3, 256, 512)
        20 is relu
        21 is conv4_2 (3, 3, 512, 512)
        22 is relu
        23 is conv4_3 (3, 3, 512, 512)
        24 is relu
        25 is conv4_4 (3, 3, 512, 512)
        26 is relu
        27 is maxpool
        28 is conv5_1 (3, 3, 512, 512)
        29 is relu
        30 is conv5_2 (3, 3, 512, 512)
        31 is relu
        32 is conv5_3 (3, 3, 512, 512)
        33 is relu
        34 is conv5_4 (3, 3, 512, 512)
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            35 is relu
            36 is maxpool
            37 is fullyconnected (7, 7, 512, 4096)
            38 is relu
            39 is fullyconnected (1, 1, 4096, 4096)
            40 is relu
            41 is fullyconnected (1, 1, 4096, 1000)
            42 is softmax
        """

        vgg = scipy.io.loadmat(path)

        vgg_layers = vgg['layers']

        def _weights(layer, expected_layer_name):
            """
            Return the weights and bias from the VGG model for a given layer.
            """
            wb = vgg_layers[0][layer][0][0][2]
            W = wb[0][0]
            b = wb[0][1]
            layer_name = vgg_layers[0][layer][0][0][0][0]
            assert layer_name == expected_layer_name
            return W, b

            return W, b

        def _relu(conv2d_layer):
            """
            Return the RELU function wrapped over a TensorFlow layer. Expects a
            Conv2d layer input.
            """
            return tf.nn.relu(conv2d_layer)

        def _conv2d(prev_layer, layer, layer_name):
            """
            Return the Conv2D layer using the weights, biases from the VGG
            model at 'layer'.
            """
            W, b = _weights(layer, layer_name)
            W = tf.constant(W)
            b = tf.constant(np.reshape(b, (b.size)))
            return tf.nn.conv2d(prev_layer, filter=W, strides=[1, 1, 1, 1], padding='SAME') + b

        def _conv2d_relu(prev_layer, layer, layer_name):
            """
            Return the Conv2D + RELU layer using the weights, biases from the VGG
            model at 'layer'.
            """
            return _relu(_conv2d(prev_layer, layer, layer_name))

        def _avgpool(prev_layer):
            """
            Return the AveragePooling layer.
            """
            return tf.nn.avg_pool(prev_layer, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1], padding='SAM

        # Constructs the graph model.
        graph = {}
        graph['input']   = tf.Variable(np.zeros((1, CONFIG.IMAGE_HEIGHT, CONFIG.IMAGE_WIDTH, CONFIG.(
        graph['conv1_1']  = _conv2d_relu(graph['input'], 0, 'conv1_1')
        graph['conv1_2']  = _conv2d_relu(graph['conv1_1'], 2, 'conv1_2')
        graph['avgpool1'] = _avgpool(graph['conv1_2'])
        graph['conv2_1']  = _conv2d_relu(graph['avgpool1'], 5, 'conv2_1')
        graph['conv2_2']  = _conv2d_relu(graph['conv2_1'], 7, 'conv2_2')
        graph['avgpool2'] = _avgpool(graph['conv2_2'])
        graph['conv3_1']  = _conv2d_relu(graph['avgpool2'], 10, 'conv3_1')
        graph['conv3_2']  = _conv2d_relu(graph['conv3_1'], 12, 'conv3_2')
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137             graph['conv3_3']  = _conv2d_relu(graph['conv3_2'], 14, 'conv3_3')
138             graph['conv3_4']  = _conv2d_relu(graph['conv3_3'], 16, 'conv3_4')
139             graph['avgpool3'] = _avgpool(graph['conv3_4'])
140             graph['conv4_1']  = _conv2d_relu(graph['avgpool3'], 19, 'conv4_1')
141             graph['conv4_2']  = _conv2d_relu(graph['conv4_1'], 21, 'conv4_2')
142             graph['conv4_3']  = _conv2d_relu(graph['conv4_2'], 23, 'conv4_3')
143             graph['conv4_4']  = _conv2d_relu(graph['conv4_3'], 25, 'conv4_4')
144             graph['avgpool4'] = _avgpool(graph['conv4_4'])
145             graph['conv5_1']  = _conv2d_relu(graph['avgpool4'], 28, 'conv5_1')
146             graph['conv5_2']  = _conv2d_relu(graph['conv5_1'], 30, 'conv5_2')
147             graph['conv5_3']  = _conv2d_relu(graph['conv5_2'], 32, 'conv5_3')
148             graph['conv5_4']  = _conv2d_relu(graph['conv5_3'], 34, 'conv5_4')
149             graph['avgpool5'] = _avgpool(graph['conv5_4'])
150 
151             return graph
152 
153         def generate_noise_image(content_image, noise_ratio = CONFIG.NOISE_RATIO):
154             """
155             Generates a noisy image by adding random noise to the content_image
156             """
157 
158             # Generate a random noise_image
159             noise_image = np.random.uniform(-20, 20, (1, CONFIG.IMAGE_HEIGHT, CONFIG.IMAGE_WIDTH, CONFIG
160 
161             # Set the input_image to be a weighted average of the content_image and a noise_image
162             input_image = noise_image * noise_ratio + content_image * (1 - noise_ratio)
163 
164             return input_image
165 
166 
167         def reshape_and_normalize_image(image):
168             """
169             Reshape and normalize the input image (content or style)
170             """
171 
172             # Reshape image to mach expected input of VGG16
173             image = np.reshape(image, ((1,) + image.shape))
174 
175             # Substract the mean to match the expected input of VGG16
176             image = image - CONFIG.MEANS
177 
178             return image
179 
180 
181         def save_image(path, image):
182 
183             # Un-normalize the image so that it looks good
184             image = image + CONFIG.MEANS
185 
186             # Clip and Save the image
187             image = np.clip(image[0], 0, 255).astype('uint8')
188             scipy.misc.imsave(path, image)
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