

```
In [1]: 1 import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
2 import numpy as np
3 import os
4 from keras import backend as K
5 from keras.preprocessing.image import load_img, save_img, img_to_array
6 import matplotlib.pyplot as plt
7 from keras.applications import vgg19
8 from keras.models import Model
9 #from keras import optimizers
10 from scipy.optimize import fmin_l_bfgs_b
11 #from keras.applications.vgg19 import VGG19
12 #vgg19_weights = '../input/vgg19/vgg19_weights_tf_dim_ordering_tf_kernels_notop.h5'
13 #vgg19 = VGG19(include_top = False, weights=vgg19_weights)
14 print(os.listdir("../input"))
15
16 # Any results you write to the current directory are saved as output.
```

Using TensorFlow backend.

```
['image-classification', 'best-artworks-of-all-time', 'vgg19']
```

```
In [2]: 1 StylePath = '../input/best-artworks-of-all-time/images/images/'
2 ContentPath = '../input/image-classification/validation/validation/travel and adventure/'
```

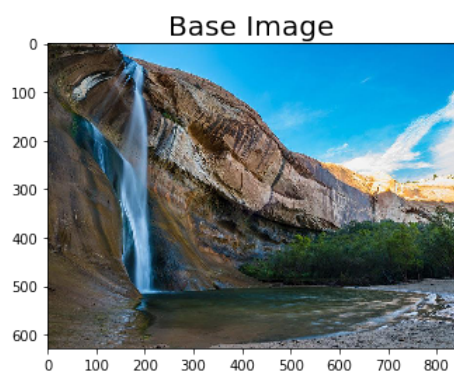
```
In [3]: 1 base_image_path = ContentPath+'13.jpg'
2 style_image_path = StylePath+'Pablo_Picasso/Pablo_Picasso_92.jpg'
```

```
In [4]: 1 # dimensions of the generated picture.
2 width, height = load_img(base_image_path).size
3 img_nrows = 400
4 img_ncols = int(width * img_nrows / height)
```

```
In [5]: 1 def preprocess_image(image_path):
2     from keras.applications import vgg19
3     img = load_img(image_path, target_size=(img_nrows, img_ncols))
4     img = img_to_array(img)
5     img = np.expand_dims(img, axis=0)
6     img = vgg19.preprocess_input(img)
7     return img
```

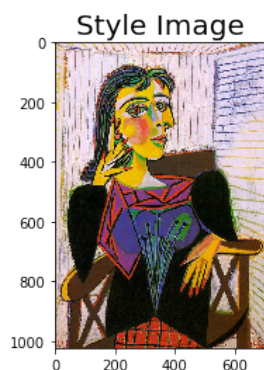
```
In [6]: 1 plt.figure()
2 plt.title("Base Image", fontsize=20)
3 img1 = load_img(ContentPath+'13.jpg')
4 plt.imshow(img1)
```

Out[6]: <matplotlib.image.AxesImage at 0x7f3814618f60>



```
In [7]: 1 plt.figure()
2 plt.title("Style Image", fontsize=20)
3 img1 = load_img(StylePath+'Pablo_Picasso/Pablo_Picasso_92.jpg')
4 plt.imshow(img1)
```

Out[7]: <matplotlib.image.AxesImage at 0x7f3811dae198>



```
In [8]: 1 # get tensor representations of our images
2
3 base_image = K.variable(preprocess_image(base_image_path))
4 style_reference_image = K.variable(preprocess_image(style_image_path))
```

WARNING:tensorflow:From /opt/conda/lib/python3.6/site-packages/tensorflow/python/framework/op_def_library.py:263: colocate_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version. Instructions for updating: Colocations handled automatically by placer.

```
In [9]: 1 K.image_data_format()
```

Out[9]: 'channels_last'

```
In [10]: 1 # this will contain our generated image
2 if K.image_data_format() == 'channels_first':
3     combination_image = K.placeholder((1,3,img_nrows, img_ncols))
4 else:
5     combination_image = K.placeholder((1,img_nrows, img_ncols,3))
```

```
In [11]: 1 # combine the 3 images into a single Keras tensor
2 input_tensor = K.concatenate([base_image,
3                               style_reference_image,
4                               combination_image
5                               ], axis=0)
```

```
In [12]: 1 # build the VGG19 network with our 3 images as input
2 # the model will be loaded with pre-trained ImageNet weights
3 from keras.applications.vgg19 import VGG19
4 vgg19_weights = '../input/vgg19/vgg19_weights_tf_dim_ordering_tf_kernels_notop.h5'
5 model = VGG19(input_tensor=input_tensor,
6               include_top=False,
7               weights=vgg19_weights)
8 #model = vgg19.VGG19(input_tensor=input_tensor,
9 #                    weights='imagenet', include_top=False)
10 print('Model loaded.')
11
```

Model loaded.

```
In [13]: 1 # Content Layer where will pull our feature maps
2 content_layers = ['block5_conv2']
3
4 # Style Layer we are interested in
5 style_layers = ['block1_conv1',
6                 'block2_conv1',
7                 'block3_conv1',
8                 'block4_conv1',
9                 'block5_conv1'
10                ]
11
12 num_content_layers = len(content_layers)
13 num_style_layers = len(style_layers)
```

```
In [14]: 1 outputs_dict = dict([(layer.name, layer.output) for layer in model.layers])
2 print(outputs_dict['block5_conv2'])
```

Tensor("block5_conv2/Relu:0", shape=(3, 25, 33, 512), dtype=float32)

```
In [15]: 1 # an auxiliary loss function
2 # designed to maintain the "content" of the
3 # base image in the generated image
4 def get_content_loss(base_content, target):
5     return K.sum(K.square(target - base_content))
```

```
In [16]: 1 import tensorflow as tf
2 # the gram matrix of an image tensor (feature-wise outer product)
3 def gram_matrix(input_tensor):
4     assert K.ndim(input_tensor)==3
5     #if K.image_data_format() == 'channels_first':
6     #     features = K.batch_flatten(input_tensor)
7     #else:
8     #     features = K.batch_flatten(K.permute_dimensions(input_tensor, (2,0,1)))
9     #gram = K.dot(features, K.transpose(features))
10    channels = int(input_tensor.shape[-1])
11    a = tf.reshape(input_tensor, [-1, channels])
12    n = tf.shape(a)[0]
13    gram = tf.matmul(a, a, transpose_a=True)
14    return gram#/tf.cast(n, tf.float32)
15
16 def get_style_loss(style, combination):
17     assert K.ndim(style) == 3
18     assert K.ndim(combination) == 3
19     S = gram_matrix(style)
20     C = gram_matrix(combination)
21     channels = 3
22     size = img_nrows*img_ncols
23     return K.sum(K.square(S - C))/((4.0 * (channels ** 2) * (size ** 2)))
24
```

```
In [17]: 1 # Get output layers corresponding to style and content layers
2 #style_outputs = [model.get_layer(name).output for name in style_layers]
3 #content_outputs = [model.get_layer(name).output for name in content_layers]
4 #model_outputs = style_outputs + content_outputs
```

```
In [18]: 1 # Get the style and content feature representations from our model
2 #style_features = [style_layer[0] for style_layer in model_outputs[:num_style_layers]]
3 #content_features = [content_layer[1] for content_layer in model_outputs[num_style_layers:]]
```

```
In [19]: 1 #gram_style_features = [gram_matrix(style_feature) for style_feature in style_features]
```

```
In [20]: 1 #style_output_features = model_outputs[:num_style_layers]
2 #content_output_features = model_outputs[num_style_layers:]
3 # Accumulate style losses from all layers
4 # Here, we equally weight each contribution of each loss layer
5 #weight_per_style_layer = 1.0 / float(num_style_layers)
6 #loss = K.variable(0.0)
7 #style_score = 0
8 #content_score = 0
9
10 #for target_style, comb_style in zip(gram_style_features, style_output_features):
11 #     style_score += weight_per_style_layer * get_style_loss(comb_style[0], target_style)
12 # Accumulate content losses from all layers
13 #weight_per_content_layer = 1.0 / float(num_content_layers)
14 #for target_content, comb_content in zip(content_features, content_output_features):
15 #     content_score += weight_per_content_layer* get_content_loss(comb_content[0], target_content)
16
17 #style_score *= style_weight
18 #content_score *= content_weight
19
20 # Get total loss
21 #loss = style_score + content_score
```

```
In [21]: 1 content_weight=0.025
2 style_weight=1.0
3 # combine these loss functions into a single scalar
4 loss = K.variable(0.0)
5 layer_features = outputs_dict['block5_conv2']
6 base_image_features = layer_features[0, :, :, :]
7 combination_features = layer_features[2, :, :, :]
8 print('Layer Feature for Content Layers :: '+str(layer_features))
9 print('Base Image Feature :: '+str(base_image_features))
10 print('Combination Image Feature for Content Layers:: '+str(combination_features)+'\n')
11 loss += content_weight * get_content_loss(base_image_features,
12                                           combination_features)
13
14 feature_layers = ['block1_conv1', 'block2_conv1',
15                  'block3_conv1', 'block4_conv1',
16                  'block5_conv1']
17 for layer_name in feature_layers:
18     layer_features = outputs_dict[layer_name]
19     style_reference_features = layer_features[1, :, :, :]
20     combination_features = layer_features[2, :, :, :]
21     print('Layer Feature for Style Layers :: '+str(layer_features))
22     print('Style Image Feature :: '+str(style_reference_features))
23     print('Combination Image Feature for Style Layers:: '+str(combination_features)+'\n')
24     s1 = get_style_loss(style_reference_features, combination_features)
25     loss += (style_weight / len(feature_layers)) * s1
26
```

Layer Feature for Content Layers :: Tensor("block5_conv2/Relu:0", shape=(3, 25, 33, 512), dtype=float32)

Base Image Feature :: Tensor("strided_slice:0", shape=(25, 33, 512), dtype=float32)

Combination Image Feature for Content Layers:: Tensor("strided_slice_1:0", shape=(25, 33, 512), dtype=float32)

WARNING:tensorflow:Variable += will be deprecated. Use variable.assign_add if you want assignment to the variable value or 'x = x + y' if you want a new python Tensor object.

Layer Feature for Style Layers :: Tensor("block1_conv1/Relu:0", shape=(3, 400, 535, 64), dtype=float32)

Style Image Feature :: Tensor("strided_slice_2:0", shape=(400, 535, 64), dtype=float32)

Combination Image Feature for Style Layers:: Tensor("strided_slice_3:0", shape=(400, 535, 64), dtype=float32)

Layer Feature for Style Layers :: Tensor("block2_conv1/Relu:0", shape=(3, 200, 267, 128), dtype=float32)

Style Image Feature :: Tensor("strided_slice_6:0", shape=(200, 267, 128), dtype=float32)

Combination Image Feature for Style Layers:: Tensor("strided_slice_7:0", shape=(200, 267, 128), dtype=float32)

Layer Feature for Style Layers :: Tensor("block3_conv1/Relu:0", shape=(3, 100, 133, 256), dtype=float32)

Style Image Feature :: Tensor("strided_slice_10:0", shape=(100, 133, 256), dtype=float32)

Combination Image Feature for Style Layers:: Tensor("strided_slice_11:0", shape=(100, 133, 256), dtype=float32)

Layer Feature for Style Layers :: Tensor("block4_conv1/Relu:0", shape=(3, 50, 66, 512), dtype=float32)

Style Image Feature :: Tensor("strided_slice_14:0", shape=(50, 66, 512), dtype=float32)

Combination Image Feature for Style Layers:: Tensor("strided_slice_15:0", shape=(50, 66, 512), dtype=float32)

Layer Feature for Style Layers :: Tensor("block5_conv1/Relu:0", shape=(3, 25, 33, 512), dtype=float32)

Style Image Feature :: Tensor("strided_slice_18:0", shape=(25, 33, 512), dtype=float32)

Combination Image Feature for Style Layers:: Tensor("strided_slice_19:0", shape=(25, 33, 512), dtype=float32)

```
In [22]: 1 def deprocess_image(x):
2     if K.image_data_format() == 'channels_first':
3         x = x.reshape((3, img_nrows, img_ncols))
4         x = x.transpose((1, 2, 0))
5     else:
6         x = x.reshape((img_nrows, img_ncols, 3))
7     # Remove zero-center by mean pixel
8     x[:, :, 0] += 103.939
9     x[:, :, 1] += 116.779
10    x[:, :, 2] += 123.68
11    # 'BGR' -> 'RGB'
12    x = x[:, :, ::-1]
13    x = np.clip(x, 0, 255).astype('uint8')
14    return x
```

```
In [23]: 1 # get the gradients of the generated image wrt the loss
2 grads = K.gradients(loss, combination_image)
3 grads
```

Out[23]: [<tf.Tensor 'gradients/concat_grad/Slice_2:0' shape=(1, 400, 535, 3) dtype=float32>]

```
In [24]: 1 outputs = [loss]
2 if isinstance(grads, (list,tuple)):
3     outputs += grads
4 else:
5     outputs.append(grads)
6 f_outputs = K.function([combination_image], outputs)
7 f_outputs
```

Out[24]: <keras.backend.tensorflow_backend.Function at 0x7f380b4c4b00>

```
In [25]: 1 # run scipy-based optimization (L-BFGS) over the pixels of the generated image
2 # so as to minimize the neural style loss
3 x_opt = preprocess_image(base_image_path)
```

```
In [26]: 1 def eval_loss_and_grads(x):
2     if K.image_data_format() == 'channels_first':
3         x = x.reshape((1, 3, img_nrows, img_ncols))
4     else:
5         x = x.reshape((1, img_nrows, img_ncols, 3))
6     outs = f_outputs([x])
7     loss_value = outs[0]
8     if len(outs[1:]) == 1:
9         grad_values = outs[1].flatten().astype('float64')
10    else:
11        grad_values = np.array(outs[1:]).flatten().astype('float64')
12    return loss_value, grad_values
13
```

```
In [27]: 1 class Evaluator(object):
2
3     def __init__(self):
4         self.loss_value = None
5         self.grads_values = None
6
7     def loss(self, x):
8         assert self.loss_value is None
9         loss_value, grad_values = eval_loss_and_grads(x)
10        self.loss_value = loss_value
11        self.grad_values = grad_values
12        return self.loss_value
13
14    def grads(self, x):
15        assert self.loss_value is not None
16        grad_values = np.copy(self.grad_values)
17        self.loss_value = None
18        self.grad_values = None
19        return grad_values
```

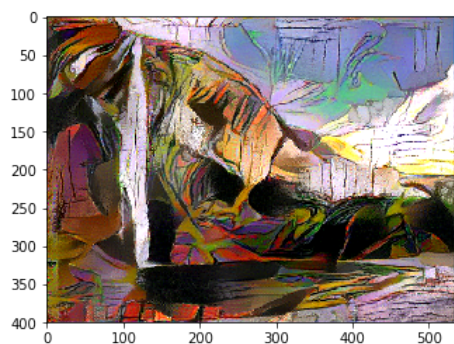
```
In [28]: 1 evaluator = Evaluator()
```

```
In [29]: 1 iterations=400
2 # Store our best result
3 best_loss, best_img = float('inf'), None
4 for i in range(iterations):
5     print('Start of iteration', i)
6     x_opt, min_val, info = fmin_l_bfgs_b(evaluator.loss,
7                                         x_opt.flatten(),
8                                         fprime=evaluator.grads,
9                                         maxfun=20,
10                                        disp=True,
11                                        )
12    print('Current loss value:', min_val)
13    if min_val < best_loss:
14        # Update best loss and best image from total loss.
15        best_loss = min_val
16        best_img = x_opt.copy()
```

```
Current loss value: 1.2474705e+20
Start of iteration 54
Current loss value: 1.23089605e+20
Start of iteration 55
Current loss value: 1.2110161e+20
Start of iteration 56
Current loss value: 1.1818676e+20
Start of iteration 57
Current loss value: 1.15673655e+20
Start of iteration 58
Current loss value: 1.1447074e+20
Start of iteration 59
Current loss value: 1.1316408e+20
Start of iteration 60
Current loss value: 1.12150336e+20
Start of iteration 61
Current loss value: 1.1121649e+20
Start of iteration 62
Current loss value: 1.0979155e+20
Start of iteration 63
```

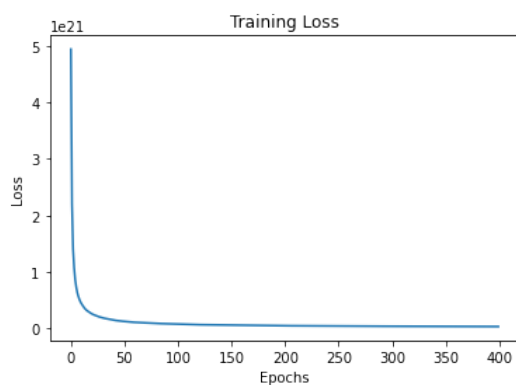
```
In [30]: 1 # save current generated image
2 imgx = deprocess_image(best_img.copy())
3 plt.imshow(imgx)
```

Out[30]: <matplotlib.image.AxesImage at 0x7f3808d83e48>



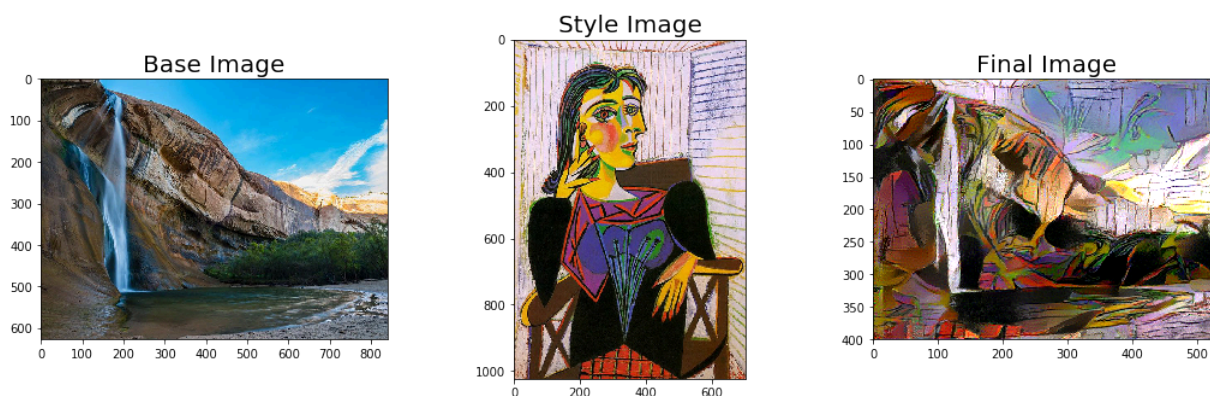
```
In [30]: 1 import numpy
2 import matplotlib.pyplot as plt
3 from graph import epochs, loss
```

```
In [33]: 1 plt.plot(epochs,loss)
2 plt.xlabel('Epochs')
3 plt.ylabel('Loss')
4 plt.title('Training Loss')
5 plt.show()
```



```
In [31]: 1 plt.figure(figsize=(30,30))
2 plt.subplot(5,5,1)
3 plt.title("Base Image",fontsize=20)
4 img_base = load_img(base_image_path)
5 plt.imshow(img_base)
6
7 plt.subplot(5,5,1+1)
8 plt.title("Style Image",fontsize=20)
9 img_style = load_img(style_image_path)
10 plt.imshow(img_style)
11
12 plt.subplot(5,5,1+2)
13 plt.title("Final Image",fontsize=20)
14 plt.imshow(imgx)
```

Out[31]: <matplotlib.image.AxesImage at 0x7f3808e045f8>



```
In [32]: 1 def preprocess_image_instantiator(image_path,img_nrows,img_ncols):
2         from keras.applications import vgg19
3         img = load_img(image_path, target_size=(img_nrows, img_ncols))
4         img = img_to_array(img)
5         img = np.expand_dims(img, axis=0)
6         img = vgg19.preprocess_input(img)
7         return img
```



```

In [33]: 1 def Run_StyleTransfer(base_image_path, style_image_path):
2
3     width, height = load_img(base_image_path).size
4     img_nrows = 400
5     img_ncols = int(width * img_nrows / height)
6
7     base_image = K.variable(preprocess_image_instantiator(base_image_path, img_nrows, img_ncols))
8     style_reference_image = K.variable(preprocess_image_instantiator(style_image_path, img_nrows, img_ncols))
9
10    if K.image_data_format() == 'channels_first':
11        combination_image = K.placeholder((1, 3, img_nrows, img_ncols))
12    else:
13        combination_image = K.placeholder((1, img_nrows, img_ncols, 3))
14
15    input_tensor = K.concatenate([base_image,
16                                  style_reference_image,
17                                  combination_image
18                                  ], axis=0)
19    from keras.applications.vgg19 import VGG19
20    vgg19_weights = '../input/vgg19/vgg19_weights_tf_dim_ordering_tf_kernels_notop.h5'
21    model = VGG19(input_tensor=input_tensor,
22                  include_top=False,
23                  weights=vgg19_weights)
24    outputs_dict = dict([(layer.name, layer.output) for layer in model.layers])
25
26    content_weight=0.025
27    style_weight=1.0
28    # combine these loss functions into a single scalar
29    loss = K.variable(0.0)
30    layer_features = outputs_dict['block5_conv2']
31    base_image_features = layer_features[0, :, :, :]
32    combination_features = layer_features[2, :, :, :]
33    #print('Layer Feature for Content Layers :: '+str(layer_features))
34    #print('Base Image Feature :: '+str(base_image_features))
35    #print('Combination Image Feature for Content Layers:: '+str(combination_image_features))
36    loss += content_weight * get_content_loss(base_image_features,
37                                              combination_features)
38
39    feature_layers = ['block1_conv1', 'block2_conv1',
40                      'block3_conv1', 'block4_conv1',
41                      'block5_conv1']
42    for layer_name in feature_layers:
43        layer_features = outputs_dict[layer_name]
44        style_reference_features = layer_features[1, :, :, :]
45        combination_features = layer_features[2, :, :, :]
46        #print('Layer Feature for Style Layers :: '+str(layer_features))
47        #print('Style Image Feature :: '+str(style_reference_features))
48        #print('Combination Image Feature for Style Layers:: '+str(combination_features))
49        sl = get_style_loss(style_reference_features, combination_features)
50        loss += (style_weight / len(feature_layers)) * sl
51
52    grads = K.gradients(loss, combination_image)
53
54    outputs = [loss]
55    if isinstance(grads, (list, tuple)):
56        outputs += grads
57    else:
58        outputs.append(grads)
59    f_outputs = K.function([combination_image], outputs)
60
61    x_opt = preprocess_image(base_image_path)
62
63    evaluator = Evaluator()
64    iterations=200
65    # Store our best result
66    best_loss, best_img = float('inf'), None
67    for i in range(iterations):
68        #print('Start of iteration', i)
69        x_opt, min_val, info = fmin_l_bfgs_b(evaluator.loss,
70                                             x_opt.flatten(),
71                                             fprime=evaluator.grads,
72                                             maxfun=20,
73                                             disp=True,
74                                             )
75        #print('Current loss value:', min_val)
76        if min_val < best_loss:
77            # Update best loss and best image from total loss.
78            best_loss = min_val
79            best_img = x_opt.copy()
80    imgx = deprocess_image(best_img.copy())
81
82    return imgx

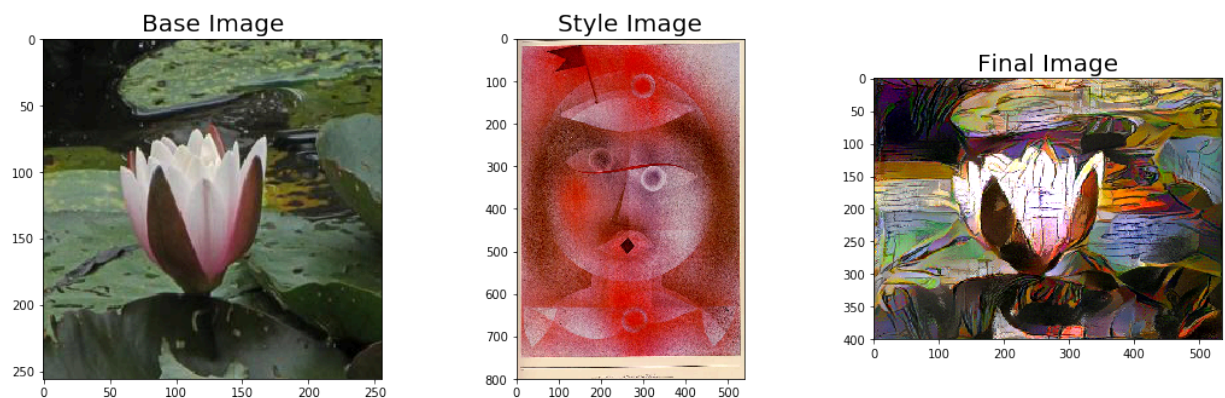
```



```
In [34]: 1 base_image_path_1 = '../input/image-classification/images/images/travel and
          2 adventure/Places365_val_00005821.jpg'
          3 plt.figure(figsize=(30,30))
          4 plt.subplot(5,5,1)
          5 plt.title("Base Image",fontsize=20)
          6 img_base = load_img(base_image_path_1)
          7 plt.imshow(img_base)
          8
          9 style_image_path_1 = '../input/best-artworks-of-all-time/images/images/Paul_Klee/Paul_Klee_96.jpg'
          10 plt.subplot(5,5,1+1)
          11 plt.title("Style Image",fontsize=20)
          12 img_style = load_img(style_image_path_1)
          13 plt.imshow(img_style)
          14
          15 plt.subplot(5,5,1+2)
          16 imgg = Run_StyleTransfer(base_image_path_1, style_image_path_1)
          17 plt.title("Final Image",fontsize=20)
          18 plt.imshow(imgg)
```

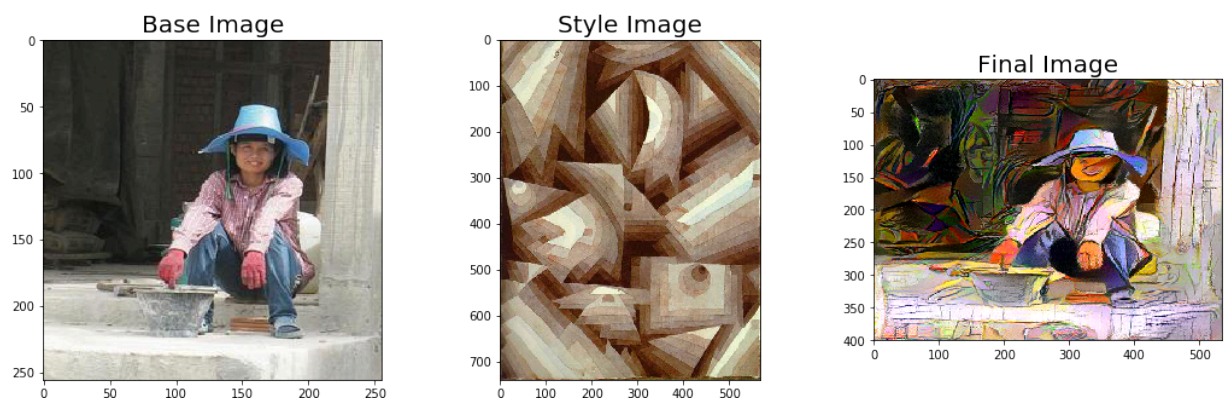
WARNING:tensorflow:Variable += will be deprecated. Use variable.assign_add if you want assignment to the variable value or 'x = x + y' if you want a new python Tensor object.

Out[34]: <matplotlib.image.AxesImage at 0x7f380b113978>



```
In [35]: 1 base_image_path_2 = '../input/image-classification/images/images/travel and
          2 adventure/Places365_val_00005982.jpg'
          3 plt.figure(figsize=(30,30))
          4 plt.subplot(5,5,1)
          5 plt.title("Base Image",fontsize=20)
          6 img_base = load_img(base_image_path_2)
          7 plt.imshow(img_base)
          8
          9 style_image_path_2 = '../input/best-artworks-of-all-time/images/images/Paul_Klee/Paul_Klee_24.jpg'
          10 plt.subplot(5,5,1+1)
          11 plt.title("Style Image",fontsize=20)
          12 img_style = load_img(style_image_path_2)
          13 plt.imshow(img_style)
          14
          15 plt.subplot(5,5,1+2)
          16 imga = Run_StyleTransfer(base_image_path_2, style_image_path_2)
          17 plt.title("Final Image",fontsize=20)
          18 plt.imshow(imga)
```

Out[35]: <matplotlib.image.AxesImage at 0x7f380a946898>

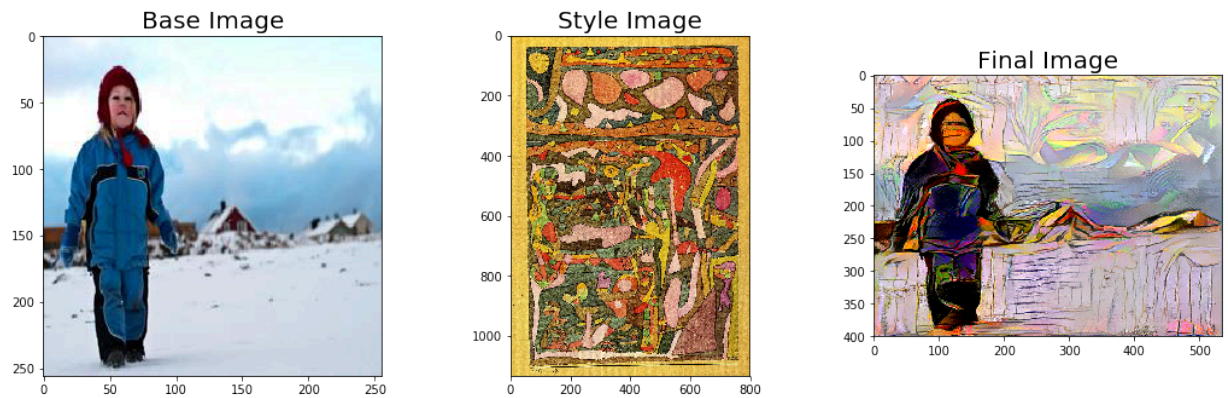


```

In [36]: 1 base_image_path_3 = '../input/image-classification/images/images/travel and
          2 adventure/Places365_val_00005752.jpg'
          3 plt.figure(figsize=(30,30))
          4 plt.subplot(5,5,1)
          5 plt.title("Base Image",fontsize=20)
          6 img_base = load_img(base_image_path_3)
          7 plt.imshow(img_base)
          8
          9 style_image_path_3 = '../input/best-artworks-of-all-time/images/images/Paul_Klee/Paul_Klee_83.jpg'
          10 plt.subplot(5,5,1+1)
          11 plt.title("Style Image",fontsize=20)
          12 img_style = load_img(style_image_path_3)
          13 plt.imshow(img_style)
          14
          15 plt.subplot(5,5,1+2)
          16 imgy = Run_StyleTransfer(base_image_path_3, style_image_path_3)
          17 plt.title("Final Image",fontsize=20)
          18 plt.imshow(imgy)

```

Out[36]: <matplotlib.image.AxesImage at 0x7f380013f358>



In [37]:

1