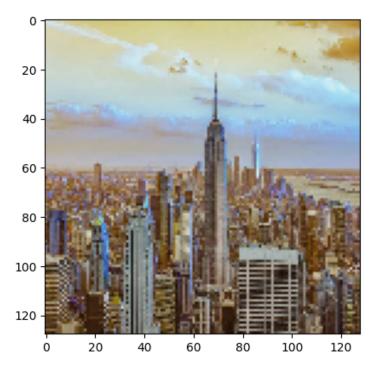
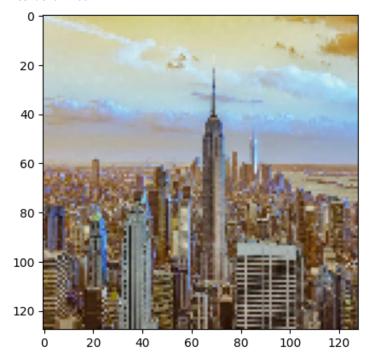
```
In [ ]: Name - Shivraj Pandurang Mane.
         Class - BE Artificial Intelligence and Data Science.
         Roll No. - 37
         Practical No. 10 - Create and Art with Neural style transfer on given image using deep learning.
 In [1]: # Import Required Libraries
 In [2]: import tensorflow as tf
         import numpy as np
         import matplotlib.pyplot as plt
         from tensorflow.keras.preprocessing import image
         from tensorflow.keras.applications import vgg19
 In [3]: # Load and Preprocess Image
 In [4]: def load_and_process_img(img_path, target_size=(128, 128)):
             img = image.load_img(img_path, target_size=target_size)
             img = image.img_to_array(img)
             img = np.expand_dims(img, axis=0)
             img = vgg19.preprocess_input(img)
             return img
         def deprocess_img(img):
             img = img.squeeze()
             img = img + [103.939, 116.779, 123.68] # Convert back to BGR
             img = np.clip(img, 0, 255).astype('uint8')
             return img
 In [5]: # Load the Content and Style Images
 In [6]: content_path = r"C:\Users\saira\Downloads\Content.jpg"
         style_path = r"C:\Users\saira\Downloads\style.jpg"
         content_img = load_and_process_img(content_path)
         style_img = load_and_process_img(style_path)
 In [7]: # Load the VGG19 Model
 In [8]: model = vgg19.VGG19(weights='imagenet', include_top=False)
         # Define the layers we will use for content and style
         content_layer = 'block5_conv2' # Content Layer
         style layers = ['block1 conv1', 'block2 conv1', 'block3 conv1', 'block4 conv1', 'block5 conv1']
 In [9]: # Define a Function to Get Feature Representations
In [10]: def get_feature_representations(model, content_img, style_img):
             content_output = model.get_layer(content_layer).output
             style outputs = [model.get layer(layer).output for layer in style layers]
             model = tf.keras.models.Model(inputs=model.input, outputs=[content_output] + style outputs)
             content_features = model(content_img)
             style_features = model(style_img)
             content_feature = content_features[0]
             style_features = style_features[1:]
             return content_feature, style_features
```

```
In [11]: # Compute the Losses (Content, Style, and Total Variation Loss)
In [12]: def compute_content_loss(content, generated):
             return tf.reduce mean(tf.square(content - generated))
         def compute style loss(style, generated):
             gram_style = gram_matrix(style)
             gram_generated = gram_matrix(generated)
             return tf.reduce_mean(tf.square(gram_style - gram_generated))
         def gram_matrix(x):
             channels = int(x.shape[-1])
             a = tf.reshape(x, [-1, channels])
             gram = tf.matmul(a, a, transpose_a=True)
             return gram
         def compute_total_variation_loss(generated):
             return tf.reduce_mean(tf.image.total_variation(generated))
In [13]: # Compute the Gradients and Perform Optimization
In [14]: def compute_loss(content_feature, style_features, generated_img):
             generated_content_feature, generated_style_features = get_feature_representations(model, generate
             content_loss = compute_content_loss(content_feature, generated_content_feature)
             style_loss = sum(compute_style_loss(style, generated) for style, generated in zip(style_features,
             total_variation_loss = compute_total_variation_loss(generated_img)
             total_loss = content_loss + 0.025 * style_loss + 1.0 * total_variation_loss
             return total loss
         def compute_gradients(content_feature, style_features, generated_img):
             with tf.GradientTape() as tape:
                 tape.watch(generated_img)
                 loss = compute_loss(content_feature, style_features, generated_img)
             grads = tape.gradient(loss, generated_img)
             return grads
In [15]: # Initialize and Optimize the Generated Image
In [16]: generated_img = tf.Variable(content_img, dtype=tf.float32)
         content_feature, style_features = get_feature_representations(model, content_img, style_img)
         optimizer = tf.optimizers.Adam(learning rate=0.01)
         iterations = 200
         for i in range(iterations):
             grads = compute_gradients(content_feature, style_features, generated_img)
             optimizer.apply_gradients([(grads, generated_img)])
             if i % 100 == 0:
                 print(f"Iteration {i}")
                 img = deprocess_img(generated_img.numpy())
                 plt.imshow(img)
                 plt.show()
```

Iteration 0

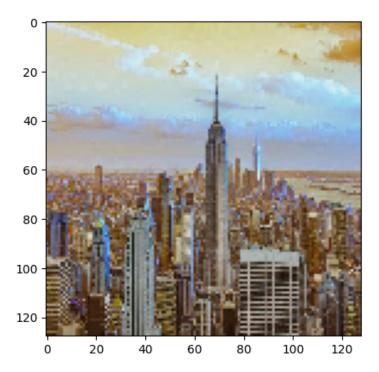


Iteration 100



In [17]: # Display the Final Image

In [18]: final_img = deprocess_img(generated_img.numpy())
 plt.imshow(final_img)
 plt.show()



In []