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### Introduction

Style transfer consists in generating an image with the same "content" as a base image, but with the "style" of a different picture (typically artistic).

#### style loss

The style loss is where the deep learning keeps in --that one is defined using a deep convolutional neural network. Precisely, it consists in a sum of L2 distances between the Gram matrices of the representations of the base image and the style reference image, extracted from different layers of a convnet (trained on ImageNet). The general idea is to capture color/texture information at different spatial scales (fairly large scales --defined by the depth of the layer considered).

#### content loss

The content loss is a L2 distance between the features of the base image (extracted from a deep layer) and the features of the combination image, keeping the generated image close enough to the original one.

#### total variation loss

The total variation loss imposes local spatial continuity between the pixels of the combination image, giving it visual coherence.

### **Description**

Transfering the style of a reference image to target image using gradient descent.



**Target image** 



Reference image

### **Description**

Transfering the style of a reference image to target image using gradient descent.



**Generated** image

### **Functions**

def preprocess\_image(image\_path):

Util function to open, resize and format pictures into appropriate tensors

def deprocess\_image(x):

Util function to convert a tensor into a valid image

### **Style transfer functions**

#### def gram\_matrix(x):

The gram matrix of an image tensor (feature-wise outer product)

#### def style\_loss(style, combination):

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 content\_loss(base, combination):

An auxiliary loss function designed to maintain the "content" of the base image in the generated image

#### def total\_variation\_loss(x):

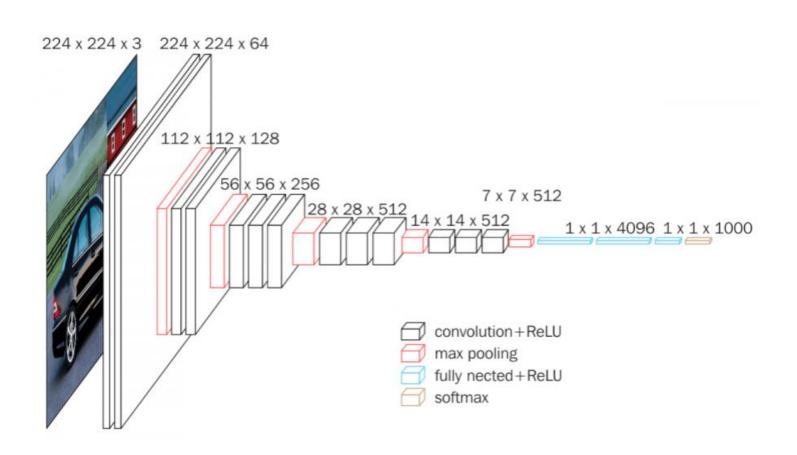
The 3rd loss function, total variation loss, designed to keep the generated image locally coherent

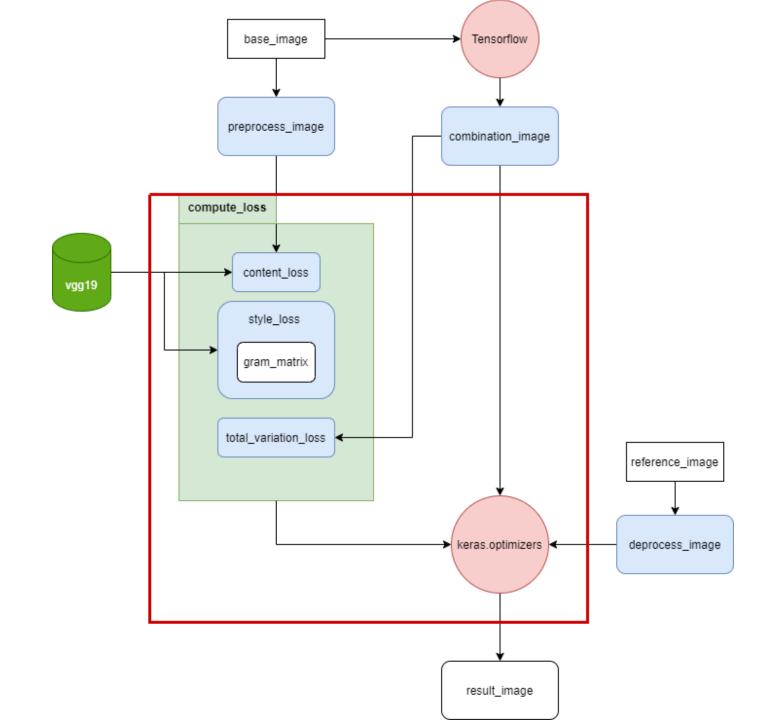
### Model

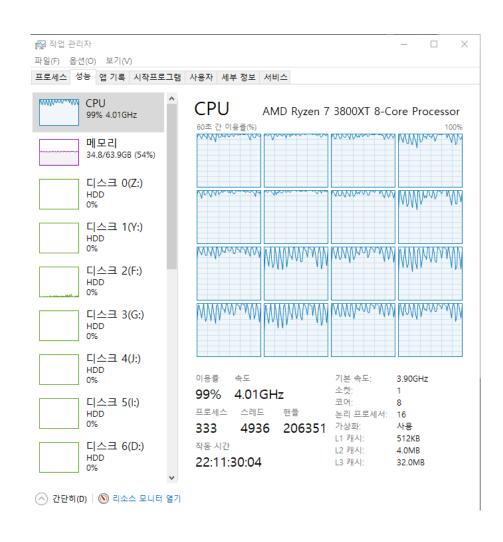
#### VGG19 algorithm or VGGNet

- VGGNet is a model developed by VGG, a research team at Oxford University, and it is the model that won the 2014 ImageNet image recognition contest.
- VGGNet means a model consisting of 16 or 19 layers.

VGG16 algorithm of VGGNet







Spec:

CPU: AMD Ryzen 7 3800XT 8Core, 16 Threads

**RAM: 64G** 

Train Mode: CPU Mode

**Train Time:** 

Start time: 21:04

End time: Next day 01:51

The Train Time: about 6 hours

```
Iteration 100: loss=2700.52
Iteration 200: loss=2140.55
Iteration 300: loss=1888.79
Iteration 400: loss=1738.28
Iteration 500: loss=1635.57
Iteration 600: loss=1559.19
Iteration 700: loss=1499.00
Iteration 800: loss=1449.74
Iteration 900: loss=1408.50
Iteration 1000: loss=1373.33
Iteration 1100: loss=1342.80
Iteration 1200: loss=1315.90
```

Iteration 3700: loss=1055.29 Iteration 3800: loss=1051.24 Iteration 3900: loss=1047.40 Iteration 4000: loss=1043.75

### **Train result**

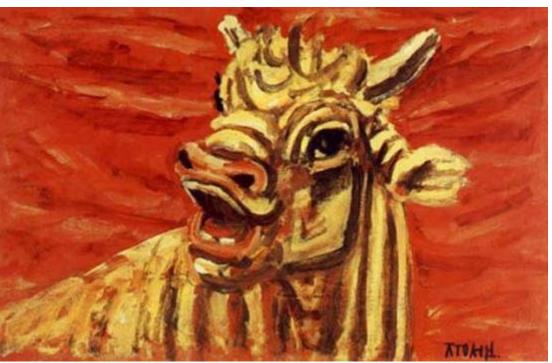




Iteration 100 Iteration 4,000

### Try other train result





**Target image** 

Reference image(Cow picture of LEE JUNG SUB)

### **Train result**



**Generated** image

Iteration 4,000

The End

Thanks!