

# Extracted the style from the image with Deep Neural Network

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## 1 Problem Description

In the field of visual art, especially painting, humans have mastered the skill to create unique visual experience through composing a complex interplay between the content and style of an image. In other area of computer vision such as object detection and recognition, Convolutional neural networks have recently enjoyed a great success in large-scale image recognition[3] which has become possible due to large public image repositories, such as ImageNet(Deng et al., 2009), and high-performance computing systems, such as GPUs or large-scale distributed clusters[1](Dean et al. 2012). In this project, we use a neural representations to separate and recombine content and style of arbitrary images. This work also offers a algorithmic understanding of how humans create and perceive artistic imagery.

## 2 The Dataset

In this project, we plan to use the datasets from ImageNet. ImageNet is a dataset of over 15 million labeled high-resolution images belong to roughly 22,000 categories. The images were collected from the web and labeled by human labelers using Amazon's Mechanical Turk crowd-sourcing tool. Starting in 2010, as part of the Pascal Visual Object Challenge, an annual competition called the ImageNet Large-Scale Visual Recognition Challenge (ILSVRC) has been held. ILSVRC uses a subset of ImageNet with roughly 1000 images in each of 1000 categories. In all, there are roughly 1.2 million training images, 50,000 validation images, and 150,000 testing images.

## 3 The Method

### 3.1 Preprocessing

Before Extracting the features from the image, we plan to rescale the image such that the shorter side was of length 256, and then cropping out the central  $256 \times 256$  patch from the resulting image.

### 3.2 Extract the features

We plan to use the convolutional neural network for extracting the features from the images. The model we plan to use is the VGG-model.

## 4 Software and hardware

### 4.1 Software

We plan to use Caffe[2](A deeplearning framework) to train our model ,use torch to apply the model to combine the style and the content from two different images.

### 4.2 Hardware

1. GPU:  
GTX-Geforce TitanX
2. CPU:  
Intel i7 4790k

## 5 Evaluation Strategy

In our project , there are two different evaluation strategy during different process. First , we use the classify testerror during training the model .During combining the style and content ,we use the loss function

$$L_{total}(\vec{p}, \vec{x}, \vec{a}) = \alpha L_{content}(\vec{p}, \vec{x}) + \beta L_{style}(\vec{x}, \vec{a}) \quad (1)$$

Where  $L_{content}$  and  $L_{style}$  is the loss between the image that we get and the original images. More over, we use this :

$$L_{content}(\vec{p}, \vec{x}, l) = \frac{1}{2} \sum_{ij} (F_{ij}^l - P_{ij}^l)^2 \quad (2)$$

## References

- [1] J. Deng, A. Berg, S. Satheesh, H. Su, A. Khosla, and Fei-Fei. Large scale visual recognition challenge. In *In http://www.image-net.org/challenges/LSVRC/2012/index*, 2012.
- [2] Y. Jia, E. Shelhamer, J. Donahue, S. Karayev, J. Long, R. Girshick, S. Guadarrama, and T. Darrell. Caffe: Convolutional architecture for fast feature embedding. *arXiv preprint arXiv:1408.5093*, 2014.
- [3] A. Krizhevsky, I. Sutskever, and G. E. Hinton. Imagenet classification with deep convolutional neural networks. *Advances in Neural Information Processing Systems*, 25:2012, 2012.