# Mosaic-Real-Time-Transfer

December 16, 2021

```
[]: !git clone https://github.com/MagicShow1999/CV-Final-Project

Cloning into 'CV-Final-Project'...
remote: Enumerating objects: 59, done.
remote: Counting objects: 100% (15/15), done.
remote: Compressing objects: 100% (13/13), done.
remote: Total 59 (delta 5), reused 9 (delta 2), pack-reused 44
Unpacking objects: 100% (59/59), done.

[]: %cd CV-Final-Project/
```

/content/CV-Final-Project

```
[]: import torch
  from torchvision import transforms, models, datasets
  from PIL import Image
  import matplotlib.pyplot as plt
  import numpy as np
  import util
  import time
  import tqdm
  import os

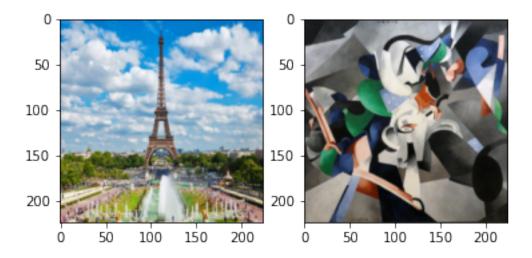
device = ("cuda" if torch.cuda.is_available() else "cpu")
```

# 1 1. Set up

```
[]: # read image
style = util.read_image('./image/udine.jpeg', 224, 224).to(device)
content = util.read_image('./image/content_img.jpg', 224, 224).to(device)

[]: # show image
fig, (ax1,ax2) = plt.subplots(1,2)
ax1.imshow(util.torchTensorToImage(content),label = "Content")
ax2.imshow(util.torchTensorToImage(style),label = "Style")
```

plt.show()



## 1.1 1.1 Hyperparmeter

## 1.2 Loss Network

```
[122]: import torch
import torch.nn as nn
from torchvision import models

class LossNet(nn.Module):
    """

    The loss network is used to define a feature reconstruction loss and a style
    →reconstruction loss
    that measure differences in content and style between images
```

```
n n n
def __init__(self):
    super(LossNet, self).__init__()
    self.vgg = models.vgg16(pretrained=True).features
    self.layers = {
        '3': "relu1_2",
        '8': "relu2 2",
        '15': "relu3_3",
        '22': "relu4 3",
    for p in self.vgg.parameters():
        p.requires_grad = False
def forward(self, x):
  features = {}
  for name,layer in self.vgg._modules.items():
    x = layer(x)
    if name in self.layers:
      features[self.layers[name]] = x
  return features
```

#### 1.3 Transform Network

#### 1.3.1 Convolution Layer

#### 1.3.2 Residual Block

```
def forward(self, x):
   identity = x
   x = self.ConvLayers(x)
   out = x + identity
   out = self.relu(out)
   return out
```

#### 1.3.3 DeConvolution Layer

```
[125]: class DeConvLayer(torch.nn.Module):
          def __init__(self, in_channels, out_channels, kernel_size, stride,_
       →upsample=None):
              super(DeConvLayer, self).__init__()
              self.upsample = upsample
              if upsample:
                   self.upsample = nn.Upsample(scale_factor=upsample, mode='nearest')
              reflection_padding = kernel_size // 2
              self.reflection_pad = nn.ReflectionPad2d(reflection_padding)
              self.conv2d = nn.Conv2d(in_channels, out_channels, kernel_size, stride)
          def forward(self, x):
              if self.upsample:
                  x = self.upsample(x)
              out = self.reflection pad(x)
              out = self.conv2d(out)
              return out
[126]: class TransformNet(torch.nn.Module):
        The image transformation network is trained to minimize a weighted \sqcup
       \rightarrow combination of loss functions
        The exact architectures of this networks can be found in the supplementary \sqcup
       \rightarrow material of the paper
        https://cs.stanford.edu/people/jcjohns/papers/fast-style/fast-style-supp.pdf
        def init (self):
            super(TransformNet, self).__init__()
            # Convolution layers
            self.ConvLayers = nn.Sequential(
                ConvLayer(3, 32, 9, 1),
                torch.nn.InstanceNorm2d(32),
                torch.nn.ReLU(),
                ConvLayer(32, 64, 3, 2),
                torch.nn.InstanceNorm2d(64),
                torch.nn.ReLU(),
                ConvLayer(64, 128, 3, 2),
```

```
torch.nn.InstanceNorm2d(128),
        torch.nn.ReLU()
    )
    # Residual blocks
    self.ResidualBlocks = nn.Sequential(
        ResidualBlock(128),
        ResidualBlock(128),
        ResidualBlock(128),
        ResidualBlock(128),
        ResidualBlock(128)
    )
    # Deconvolutional layers
    self.DeConvLayers = nn.Sequential(
        DeConvLayer(128, 64, kernel_size=3, stride=1, upsample=2),
        torch.nn.InstanceNorm2d(64, affine=True),
        DeConvLayer(64, 32, kernel_size=3, stride=1, upsample=2),
        torch.nn.InstanceNorm2d(32, affine=True),
        DeConvLayer(32, 3, kernel_size=9, stride=1),
        torch.nn.InstanceNorm2d(3, affine=True)
    )
def forward(self, x):
    x = self.ConvLayers(x)
    x = self.ResidualBlocks(x)
    x = self.DeConvLayers(x)
    return x
```

## 2 Train

## 2.0.1 2.1 Prepare Training set (COCO) TODO

Currently use val2014 instead because train2014 is too big

```
[120]: # Download dataset (COCO2017)

# https://cocodataset.org/#download

# TODO change val to train

!wget http://images.cocodataset.org/zips/val2017.zip
```

```
--2021-12-16 07:46:52-- http://images.cocodataset.org/zips/val2017.zip
Resolving images.cocodataset.org (images.cocodataset.org)... 52.217.9.188
Connecting to images.cocodataset.org
(images.cocodataset.org)|52.217.9.188|:80... connected.
HTTP request sent, awaiting response... 200 OK
```

```
Saving to: val2017.zip
     val2017.zip
                        in 49s
     2021-12-16 07:47:41 (15.9 MB/s) - val2017.zip saved [815585330/815585330]
[127]: !rm -rf dataset
      !rm -rf model checkpoints
      !rm -rf transformed_images
      !mkdir dataset
      !mkdir model_checkpoints
      !mkdir transformed_images
[128]: || unzip -qq val2017.zip -d /content/CV-Final-Project/dataset/val2017
     !rm val2017.zip
[129]: # Get Training Dataset
     transform = util.common_transforms(TRAIN_IMAGE_SIZE, TRAIN_IMAGE_SIZE)
     # ref: https://pytorch.org/vision/0.8/datasets.html#imagefolder
     train_dataset = datasets.ImageFolder("/content/CV-Final-Project/dataset/", __
      →transform=transform)
     train_loader = torch.utils.data.DataLoader(train_dataset,__
      →batch_size=BATCH_SIZE, shuffle=True)
```

## 2.0.2 2.2 Style Feature for Reference

Length: 815585330 (778M) [application/zip]

```
[131]: # Style Features
B, C, H, W = style.shape
style_features = vgg(style.expand([BATCH_SIZE, C, H, W]))
style_grams = {}
for layer, feature in style_features.items():
    style_grams[layer] = util.gram_matrix(feature)
```

#### 2.0.3 2.3 Model & Optimizer

```
[130]: # Model
transform_net = TransformNet().to(device)
vgg = LossNet().to(device)

# Optimizer
optimizer = torch.optim.Adam(transform_net.parameters(),lr=LR)
```

#### 2.0.4 2.4 Training

```
[132]: content_loss_history = []
      style_loss_history = []
      total_loss_history = []
      begin_time = time.time()
      batch_count = 0
      for epoch in range(1, EPOCHS + 1):
       content_loss_sum = 0
       style loss sum = 0
       total_loss_sum = 0
       for batch_idx, (data, _) in enumerate(train_loader):
          curr_batch_size = data.shape[0]
         batch count += 1
         if device == "cuda":
           torch.cuda.empty_cache() # Free-up memory
         optimizer.zero_grad()
         # Get transformed images and features generated by transformed images
         content_imgs = data.to(device)
         transformed_imgs = transform_net(content_imgs)
          content features = vgg(content imgs)
         transformed_features = vgg(transformed_imgs)
          # Content Loss
         MSELoss = nn.MSELoss().to(device)
          content_loss = CONTENT_WEIGHT * MSELoss(transformed_features['relu3_3'],__
       content_loss_sum += content_loss.item()
         # Style Loss
         style loss = 0
         for layer, feature in transformed_features.items():
            style gram = style grams[layer][:curr batch size]
            transformed_gram = util.gram_matrix(feature)
            style_loss += MSELoss(transformed_gram, style_gram) * STYLE_WEIGHTS[layer]
          style_loss_sum += style_loss.item()
          # Total Loss
         total_loss = content_loss + style_loss
         total_loss_sum += total_loss.item()
          # Backprop and Weight Update
         total_loss.backward()
         optimizer.step()
```

```
print('\nTrain Epoch: {} [{}/{} ({:.0f}%)]'.format(
             epoch, batch_idx * len(data), len(train_loader.dataset),
               100. * batch_idx / len(train_loader)))
        print('\tContent Loss: {:.6f}\tStyle Loss: {:.6f}\tTotal Loss: {:.6f}\'.
 →format(
             content_loss_sum/batch_count, style_loss_sum/batch_count,_
 →total loss sum/batch count))
        print("Time elapsed:\t{} seconds".format(time.time()-begin_time))
        content_loss_history.append(content_loss_sum/batch_count)
        style_loss_history.append(style_loss_sum/batch_count)
        total loss history.append(total loss sum/batch count)
        # Save model checkpoint
        model_file = '/content/CV-Final-Project/model_checkpoints/

→checkpoint_{}_{}.pth'.format(epoch, batch_idx//LOG_INTERVAL)

        torch.save(transform net.state dict(), model file)
        print('Saved model to ' + model_file + '.')
        # Save image
        transformed_img = transformed_imgs[0].clone().detach().unsqueeze(dim=0)
        image_path = '/content/CV-Final-Project/transformed_images/img_{}_{}.
 →png'.format(epoch, batch_idx*4)
        util.save image(transformed img, image path)
  # Save model weights
  final_path = "/content/CV-Final-Project/model_checkpoints/
 \rightarrowtransformation_network_weight.pth"
  torch.save(transform_net.state_dict(), final_path)
  print('Saved model to ' + final_path + '.')
Train Epoch: 1 [0/5000 (0%)]
        Content Loss: 31.052502 Style Loss: 1471.257935 Total Loss: 1502.310425
Time elapsed:
                0.2166118621826172 seconds
Saved model to /content/CV-Final-Project/model_checkpoints/checkpoint_1_0.pth.
Successfully save the final stylized image to: /content/CV-Final-
Project/transformed_images/img_1_0.png
Train Epoch: 1 [2000/5000 (40%)]
        Content Loss: 16.060682 Style Loss: 23.980723 Total Loss: 40.041405
Time elapsed:
               72.9520115852356 seconds
Saved model to /content/CV-Final-Project/model_checkpoints/checkpoint_1_1.pth.
Successfully save the final stylized image to: /content/CV-Final-
Project/transformed_images/img_1_2000.png
Train Epoch: 1 [4000/5000 (80%)]
        Content Loss: 15.210442 Style Loss: 16.709062 Total Loss: 31.919505
                145.7732515335083 seconds
Time elapsed:
Saved model to /content/CV-Final-Project/model_checkpoints/checkpoint_1_2.pth.
```

if batch\_idx % LOG\_INTERVAL == 0:

```
Successfully save the final stylized image to: /content/CV-Final-
Project/transformed_images/img_1_4000.png
Saved model to /content/CV-Final-
Project/model_checkpoints/transformation_network_weight.pth.
Train Epoch: 2 [0/5000 (0%)]
        Content Loss: 0.010151 Style Loss: 0.006639
                                                       Total Loss: 0.016790
Time elapsed:
               181.87240886688232 seconds
Saved model to /content/CV-Final-Project/model_checkpoints/checkpoint_2_0.pth.
Successfully save the final stylized image to: /content/CV-Final-
Project/transformed_images/img_2_0.png
Train Epoch: 2 [2000/5000 (40%)]
        Content Loss: 3.611654 Style Loss: 2.177996
                                                       Total Loss: 5.789650
                254.1175127029419 seconds
Saved model to /content/CV-Final-Project/model_checkpoints/checkpoint_2_1.pth.
Successfully save the final stylized image to: /content/CV-Final-
Project/transformed_images/img_2_2000.png
Train Epoch: 2 [4000/5000 (80%)]
        Content Loss: 5.449342 Style Loss: 3.277000 Total Loss: 8.726342
Time elapsed:
                327.0498912334442 seconds
Saved model to /content/CV-Final-Project/model_checkpoints/checkpoint_2_2.pth.
Successfully save the final stylized image to: /content/CV-Final-
Project/transformed_images/img_2_4000.png
Saved model to /content/CV-Final-
Project/model_checkpoints/transformation_network_weight.pth.
```

# [133]: zip -r /content/CV-Final-Project/imgs.zip /content/CV-Final-Project/ -transformed\_images

```
updating: content/CV-Final-Project/transformed_images/ (stored 0%)
updating: content/CV-Final-Project/transformed_images/img_1_4000.png (deflated 0%)
updating: content/CV-Final-Project/transformed_images/img_1_2000.png (deflated 0%)
updating: content/CV-Final-Project/transformed_images/img_1_0.png (deflated 0%)
adding: content/CV-Final-Project/transformed_images/img_2_4000.png (deflated 0%)
adding: content/CV-Final-Project/transformed_images/img_2_2000.png (deflated 0%)
adding: content/CV-Final-Project/transformed_images/img_2_0.png (deflated 0%)
```

## [134]: | ! pwd

/content/CV-Final-Project

```
[135]: cnt = len(train_dataset)//(BATCH_SIZE * LOG_INTERVAL)
    _, ax = plt.subplots(cnt, 1,figsize = (8,cnt*8))
    epoch = EPOCHS
    for i in range(1, cnt+1):
        idx = LOG_INTERVAL * i * BATCH_SIZE
        image_name = 'img_{}_{}_{}.png'.format(epoch, idx)
        image_path = './transformed_images/' + image_name
        sampe_image = util.read_image(image_path, TRAIN_IMAGE_SIZE, TRAIN_IMAGE_SIZE).
        -to(device)
        ax[i-1].set_title(image_name)
        ax[i-1].imshow(util.torchTensorToImage(sampe_image))
        ax[i-1].axis("off")
        plt.show()
```

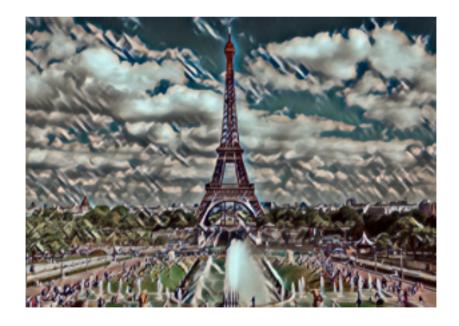
img\_2\_2000.png



img\_2\_4000.png



Time taken: 0.10517358779907227



Successfully save the final stylized image to: /content/CV-Final-Project/image/udine\_out1.jpg

Time taken: 0.520117998123169



Successfully save the final stylized image to: /content/CV-Final-Project/image/udine\_out2.jpg

```
[140]: zip -r /content/CV-Final-Project/models.zip /content/CV-Final-Project/

model_checkpoints
```

```
adding: content/CV-Final-Project/model_checkpoints/ (stored 0%)
  adding: content/CV-Final-Project/model_checkpoints/checkpoint_2_1.pth
(deflated 8%)
  adding: content/CV-Final-Project/model_checkpoints/checkpoint_1_0.pth
(deflated 9%)
  adding: content/CV-Final-
Project/model_checkpoints/transformation_network_weight.pth (deflated 8%)
  adding: content/CV-Final-Project/model_checkpoints/checkpoint_2_0.pth
(deflated 8%)
  adding: content/CV-Final-Project/model_checkpoints/checkpoint_1_1.pth
(deflated 8%)
  adding: content/CV-Final-Project/model_checkpoints/checkpoint_1_2.pth
(deflated 8%)
  adding: content/CV-Final-Project/model_checkpoints/checkpoint_2_2.pth
(deflated 8%)
```

Saving the notebook as pdf.

```
[]: !apt-get install texlive texlive-xetex texlive-latex-extra pandoc
!pip install pypandoc

from google.colab import drive
drive.mount('/content/drive')
```

```
[147]: cp /content/drive/MyDrive/"Colab Notebooks"/Mosaic-Real-Time-Transfer.ipynb ./
!jupyter nbconvert --to PDF "Mosaic-Real-Time-Transfer.ipynb"
```

```
[NbConvertApp] Converting notebook Mosaic-Real-Time-Transfer.ipynb to PDF
[NbConvertApp] Support files will be in Mosaic-Real-Time-Transfer_files/
[NbConvertApp] Making directory ./Mosaic-Real-Time-Transfer_files
[NbConvertApp] Making directory ./Mosaic-Real-Time-Transfer_files
[NbConvertApp] Making directory ./Mosaic-Real-Time-Transfer_files
[NbConvertApp] Making directory ./Mosaic-Real-Time-Transfer_files
[NbConvertApp] Writing 92715 bytes to ./notebook.tex
[NbConvertApp] Building PDF
[NbConvertApp] Running xelatex 3 times: [u'xelatex', u'./notebook.tex', '-quiet']
[NbConvertApp] Running bibtex 1 time: [u'bibtex', u'./notebook']
[NbConvertApp] WARNING | bibtex had problems, most likely because there were no citations
[NbConvertApp] PDF successfully created
[NbConvertApp] Writing 978286 bytes to Mosaic-Real-Time-Transfer.pdf
```