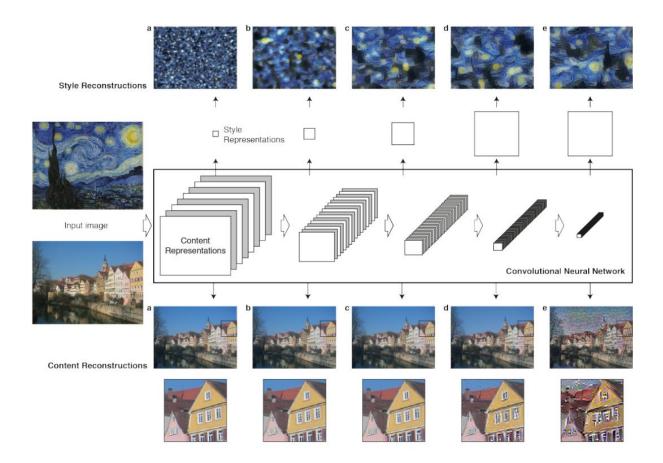
## MRP - Week 7

## This week we will be experimenting with the input layers taken to evaluate the loss function.

- Content Representation
  - o Higher layers capture high-level content features. Not exact pixel value
  - Higher layers are more concerned with the shapes and arrangement of objects in the image
  - We care about these last layers because they contain the content we want to make artistic
- Style Reconstruction
  - To extract the style of an image (i.e. the stroke and colors), we calculate the correlation between the outputs of various layers
  - o The similarities(correlations) between various layers captures information about the texture and colors



- The figure above shows a 5 layer Convolutional Network
- The top row (style reconstruction) shows the style extracted each layer
  - The style of the input image from style representations built on different subsets of CNN layers ('conv1 1' (a), 'conv1 1' and 'conv2 1' (b), 'conv1 1', 'conv2 1' and 'conv3 1' (c), 'conv1 1', 'conv2 1', 'conv3 1' and 'conv4 1' (d), 'conv1 1', 'conv2 1', 'conv3 1', 'conv4 1' and 'conv5 1' (e))

- The bottom row (content reconstruction) shows how shapes/colors are preserved but pixels are lost with deeper layers
  - We reconstruct the content image from layers 'conv1 1' (a), 'conv2 1' (b), 'conv3 1' (c), 'conv4 1' (d) and 'conv5 1' (e)
- The Layers of a Convolutional Network are given in the figure below.
- The green outline indicates the layers of VGG-16 network we will be using

ConvNet Configuration					
A	A-LRN	В	С	D	Е
11 weight	11 weight	13 weight	16 weight	16 weight	19 weight
layers	layers	layers	layers	layers	layers
input (224 × 224 RGB image)					
conv3-64	conv3-64	conv3-64	conv3-64	conv3-64	conv3-64
	LRN	conv3-64	conv3-64	conv3-64	conv3-64
maxpool					
conv3-128	conv3-128	conv3-128	conv3-128	conv3-128	conv3-128
		conv3-128	conv3-128	conv3-128	conv3-128
maxpool					
conv3-256	conv3-256	conv3-256	conv3-256	conv3-256	conv3-256
conv3-256	conv3-256	conv3-256	conv3-256	conv3-256	conv3-256
			conv1-256	conv3-256	conv3-256
					conv3-256
maxpool					
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
			conv1-512	conv3-512	conv3-512
					conv3-512
maxpool					
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
conv3-512	conv3-512	conv3-512	conv3-512	conv3-512	conv3-512
			conv1-512	conv3-512	conv3-512
					conv3-512
maxpool					
FC-4096					
FC-4096					
FC-1000					
soft-max					

```
'block1_conv1/Relu:0' shape=(3, 300, 300, 64) c
'block1 conv2/Relu:0' shape=(3, 300, 300, 64) c
'block1_pool/MaxPool:0' shape=(3, 150, 150, 64)
'block2_conv1/Relu:0' shape=(3, 150, 150, 128)
'block2_conv2/Relu:0' shape=(3, 150, 150, 128)
'block2_pool/MaxPool:0' shape=(3, 75, 75, 128) c
'block3_conv1/Relu:0' shape=(3, 75, 75, 256) dt
'block3 conv2/Relu:0' shape=(3, 75, 75, 256) dt
'block3_conv3/Relu:0' shape=(3, 75, 75, 256) dt
'block3 pool/MaxPool:0' shape=(3, 37, 37, 256) c
'block4_conv1/Relu:0' shape=(3, 37, 37, 512) dt
'block4_conv2/Relu:0' shape=(3, 37, 37, 512) dt
'block4_conv3/Relu:0' shape=(3, 37, 37, 512) dt
'block4 pool/MaxPool:0' shape=(3, 18, 18, 512) c
'block5_conv1/Relu:0' shape=(3, 18, 18, 512) dt
'block5 conv2/Relu:0' shape=(3, 18, 18, 512) dt
'block5 conv3/Relu:0' shape=(3, 18, 18, 512) dt
'block5_pool/MaxPool:0' shape=(3, 9, 9, 512) dty
```

```
Loss function L_{total}(S,C,G) = \alpha L_{content}(C,G) + \beta L_{style}(S,G)
```

- Leon Gatys used the following layers as input for the loss functions in his paper:
  - Content loss: block4 conv2
  - Style loss: block1\_conv1, block2\_conv1, block3\_conv1, block4\_conv1, block5\_conv1
  - $\circ$   $\alpha / \beta = 1x10^{-3} \text{ or } 1x10^{-4}$
- The layers used for the style loss function are spread across the all 5 blocks of the VGG-16 network. This gives the best results because we are able to form correlations between at least one layer of each block
- We can vary the layer used for the content loss function to see which layer gives the best results

## **Experiments:**

 $\alpha = 0.5$  $\beta = 5.0$ 

Style layers = block1\_conv1, block2\_conv1, block3\_conv1, block4\_conv1, block5\_conv1

Exp#1: Content layer = block1\_cov2



Exp#2: Content layer = block2\_cov2



Exp#3:
Content layer = block3\_cov3



Exp#4: Content layer = block4\_cov3



Exp#5:
Content layer = block5\_cov3



The results of the generated image before block3\_conv3 are weak and after block4\_conv3 are too strong. Maybe the ideal layer is somewhere in block 4 layer 1 or 2.

Exp#6: Content layer = block4\_cov1



Exp#7:
Content layer = block4\_cov



Experiment 6 seems to offer the best results visually speaking.