

# Deep Learning

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Materials mostly by Andrew Ng

March 21, 2018





# Outline

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Convolution  
Matrix

Parameter

Activation  
Function

Pooling  
Layer

Add Fully  
Connected  
Layer

## 1 Convolution Matrix

## 2 Parameter

## 3 Activation Function

## 4 Pooling Layer

## 5 Add Fully Connected Layer



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# What is Convolution?

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# Filter Matrix: Blur

Convolution  
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0	0	0	0	0
0	1	1	1	0
0	1	1	1	0
0	1	1	1	0
0	0	0	0	0

(a) Filter



(b) Convolved Image

Figure: Blur

# Filter Matrix: Sharpen

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0	0	0	0	0
0	0	-1	0	0
0	-1	5	-1	0
0	0	-1	0	0
0	0	0	0	0

(a) Filter



(b) Convolved Image

Figure: Sharpen



# Filter Matrix: Edge Enhance

Convolution  
Matrix

Parameter

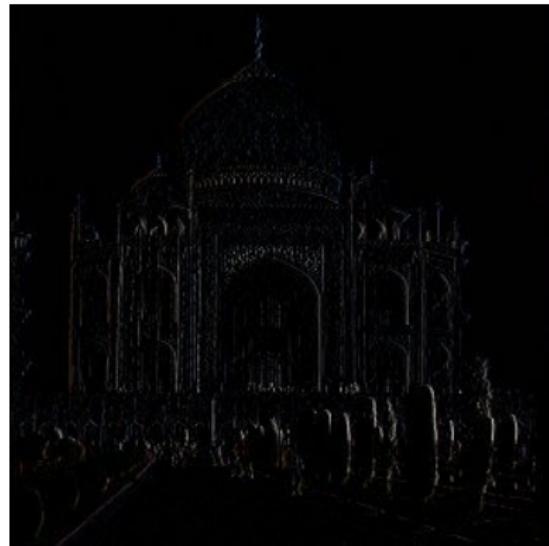
Activation  
Function

Pooling  
Layer

Add Fully  
Connected  
Layer

0	0	0	
-1	1	0	
0	0	0	

(a) Filter



(b) Convolved Image

Figure: Edge Enhance

# Filter Matrix: Edge Detect

Convolution  
Matrix

Parameter

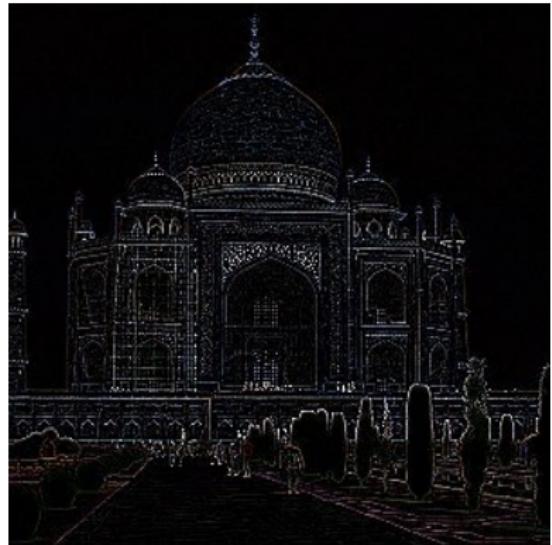
Activation  
Function

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Layer

Add Fully  
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0	1	0	
1	-4	1	
0	1	0	

(a) Filter



(b) Convolved Image

Figure: Edge Detect



# Feature Map or Convolved Image

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# Each filter result in a feature map

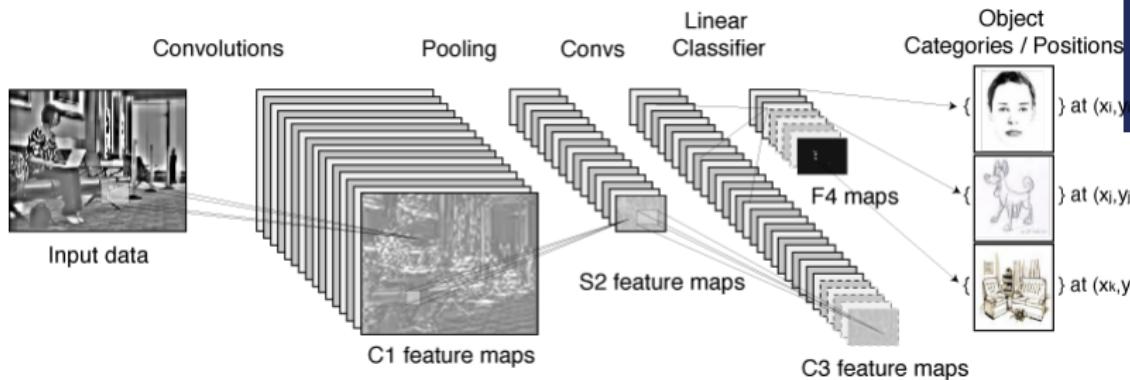
Convolution Matrix

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# Parameter

Convolution  
Matrix

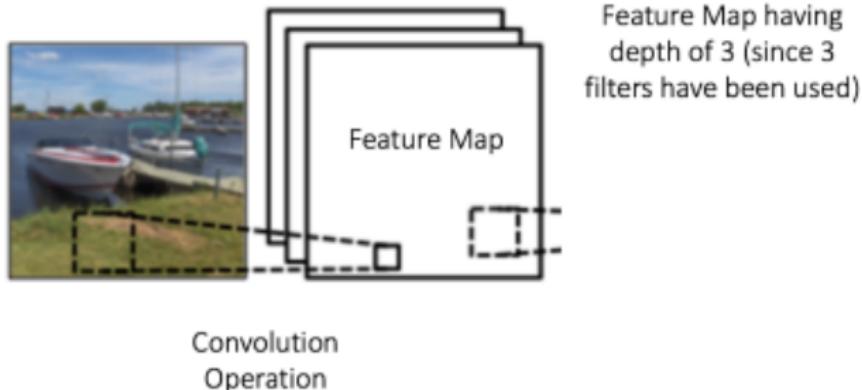
Parameter

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**Depth:** Depth corresponds to the number of filters we use for the convolution operation.





# Parameter

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**Stride:** Stride is the number of pixels by which we slide our filter matrix over the input matrix. When the stride is 2, then the filters jump 2 pixels at a time as we slide them around.



# Parameter

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**Zero-padding:** Sometimes, it is convenient to pad the input matrix with zeros around the border, so that we can apply the filter to bordering elements of our input image matrix. There are also other types of padding methods.

0	0	0	0	0	0	0	0
0							0
0							0
0							0
0							0
0							0
0							0
0							0

original 6x6

Zero Padding



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# ReLU

Convolution Matrix

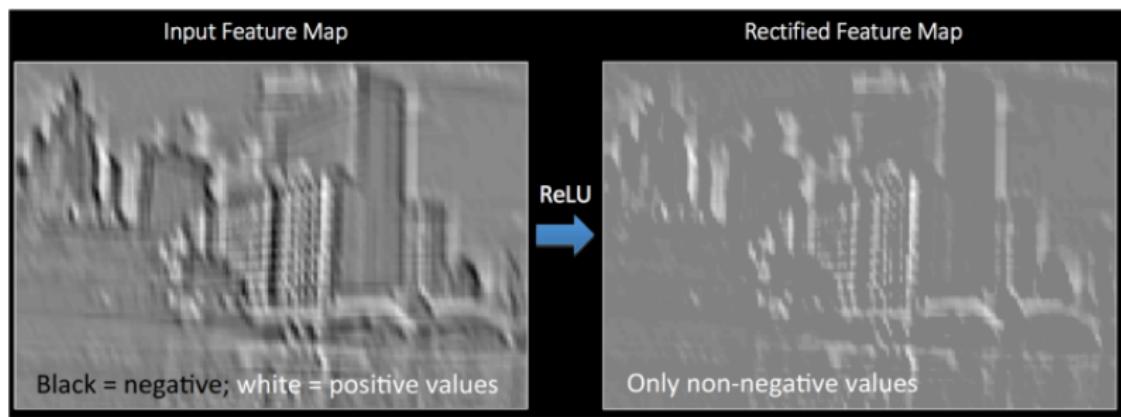
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Activation Function

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The purpose of ReLU is to introduce non-linearity in our ConvNet, since most of the real-world data we would want our ConvNet to learn would be non-linear (Convolution is a linear operation – element wise matrix multiplication and addition, so we account for non-linearity by introducing a non-linear function like ReLU).





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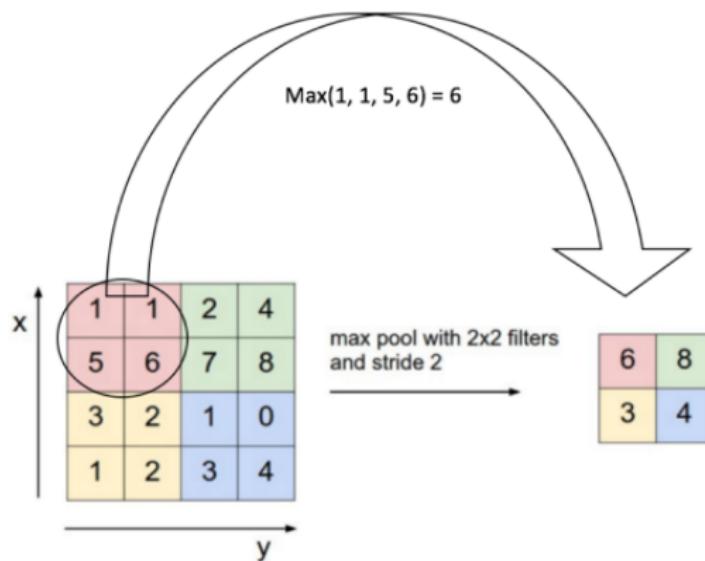
3 Activation Function

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# Pooling Layer

Spatial Pooling (also called subsampling or downsampling) reduces the dimensionality of each feature map but retains the most important information. Spatial Pooling can be of different types: Max, Average, Sum etc.





# Pooling Layer

Convolution  
Matrix

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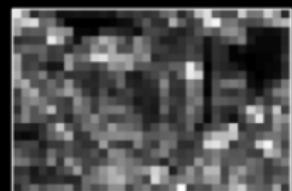
Pooling  
Layer

Add Fully  
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Layer

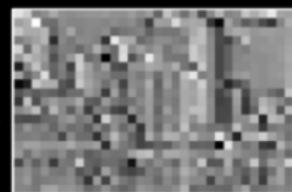


Pooling

Max



Sum





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# Add Fully Connected Layer

Convolution Matrix

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In the end, we flatten all the arrays to 1-dimension vectors and use them for training.

