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Computer Vision Topics

Image Segmentation

Image Classification

Object Localization (1 object) or Detection (several objects)

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Making Each Kernel Operation Non-Linear

$$y_0 = w_0x_0 + w_1x_1 + w_2x_2 + w_3x_4 + w_4x_5 + w_5x_6 + w_6x_8 + w_7x_9 + w_8x_{10}$$

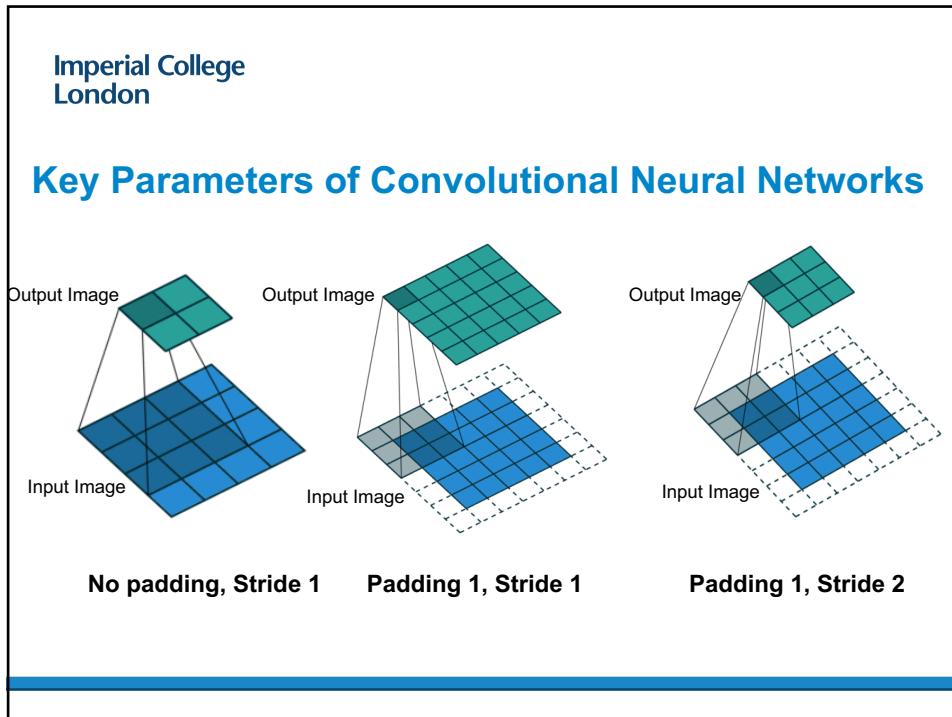
A bias term can be added

$$y_0 = w_0x_0 + w_1x_1 + w_2x_2 + w_3x_4 + w_4x_5 + w_5x_6 + w_6x_8 + w_7x_9 + w_8x_{10} + b$$

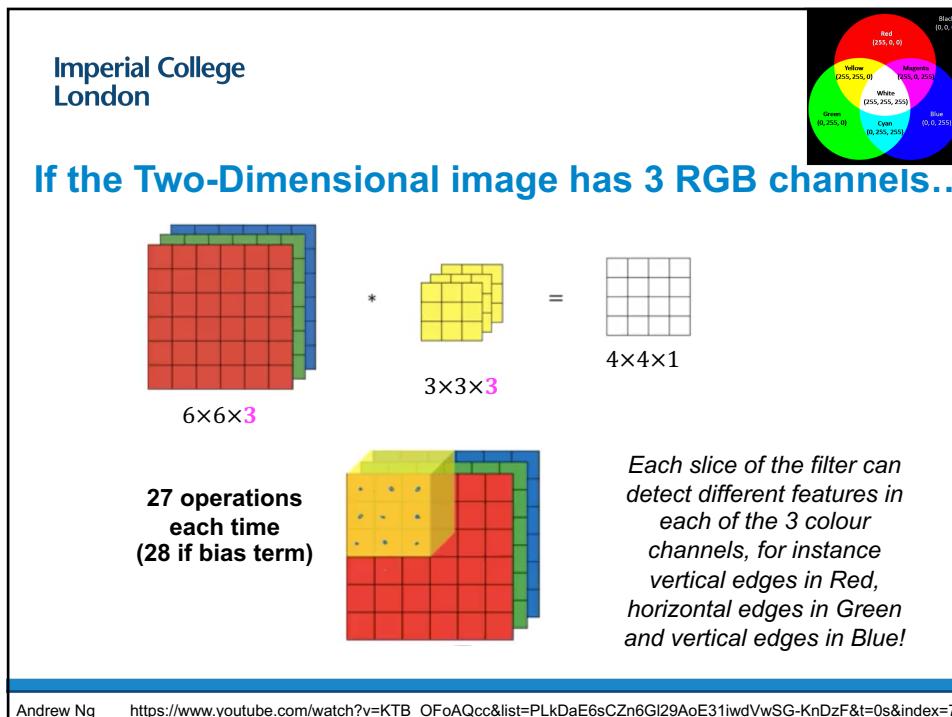
And a non-linear transformation is applied by an activation function g

$$y_0 = g(w_0x_0 + w_1x_1 + w_2x_2 + w_3x_4 + w_4x_5 + w_5x_6 + w_6x_8 + w_7x_9 + w_8x_{10} + b)$$

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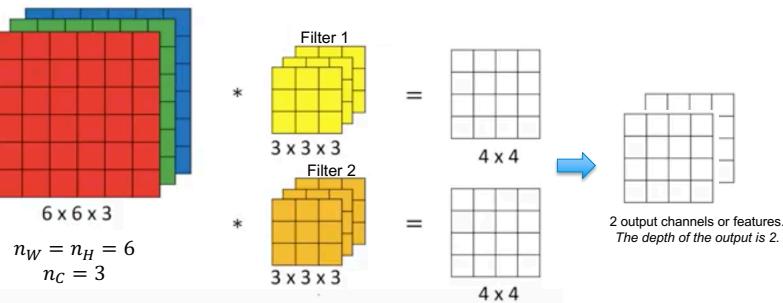
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If Different Filters are Applied ...



Exercise: if my input Image is 50×50 with 10 channels, and if I apply 25 filters each of size $3 \times 3 \times 10$ with a stride of 1, no padding, what is the size and number of channels of the output image?

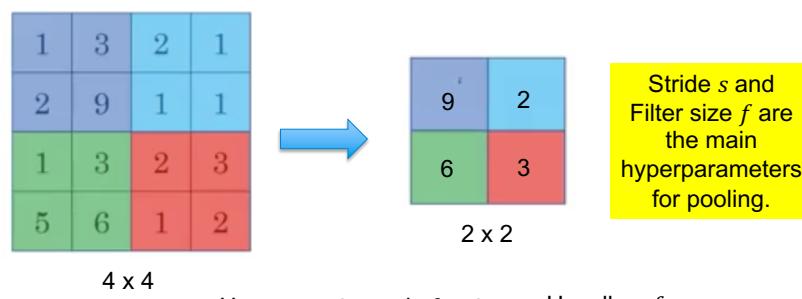
Answer: size is 48×48 . with 25 channels

Andrew Ng https://www.youtube.com/watch?v=KTB_OFoAQcc&list=PLkDaE6sCZn6Gl29AcE31iwdVwSG-KnDzF&t=0s&index=7

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Example of Max Pooling Layer (for Downsampling)



Here $s = 2$ and $f = 2$ Usually $f = s$

Pooling makes the input representations (feature dimension) smaller and more manageable for the next layer. Max Pooling is used because it may be interesting to keep the high values for the activation of the next layer as they may characterize some important features. Pooling reduces the number of parameters and computations in the network, therefore controlling overfitting.

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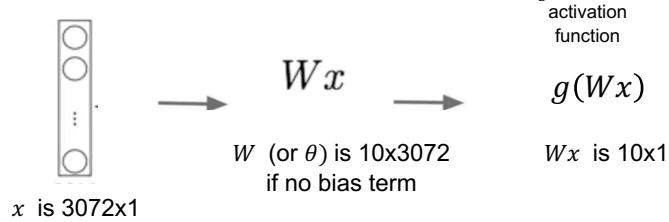
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We already know Fully Connected Layers!

Transform for example 32x32x3 image into 10x1 vector.

1. Reshape image into 1D vector x of dimension $32 \times 32 \times 3 = 3072$

2. Apply full convolution through matrix W

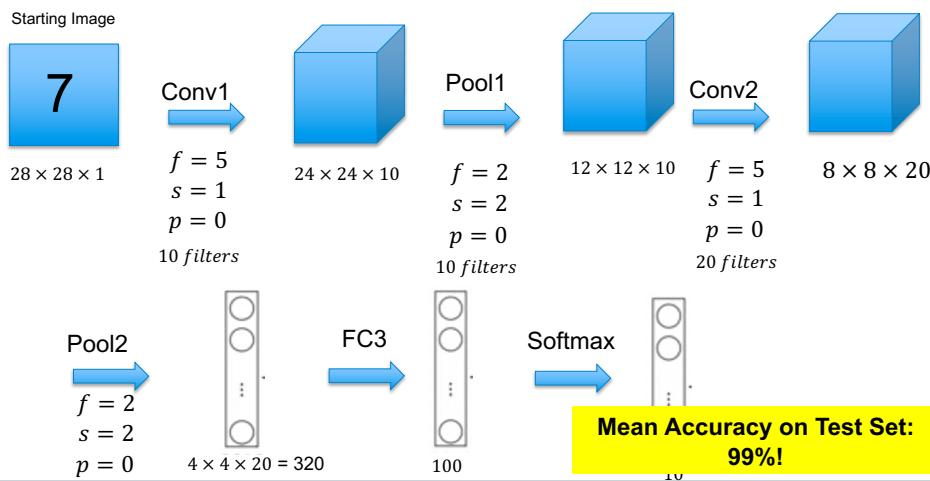


Fully connected layers often have many weights as they connect every neuron in input layer to every neuron in output layer.

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MNIST CNN Example (The LeNet-5 Architecture from LeCun)



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MNIST: Summary of CNN Layers (Convolutions with Bias Terms)										
	Size of input image n	Number of input channels	f	p	s	Size of output image $(n+2p-f)/s+1$	Number of output channels or filters	Number of output neurons	Size of Filter + 1	Number of Parameters
Conv1	28	1	5	0	1	24	10	5760	26	260
Pool1	24	10	2	0	2	12	10	1440		
Conv2	12	10	5	0	1	8	20	1280	251	5020
Pool2	8	20	2	0	2	4	20	320		
	Size of input							Number of output neurons		
FC1	320							100	321	32100
Softmax	100							10	101	1010
									Total Neurons	8910
									Total Parameters	38390

Formula used for last column:
Number of Parameters = Number of output channels × (Size of filter+1) (as we assume there is a bias term)

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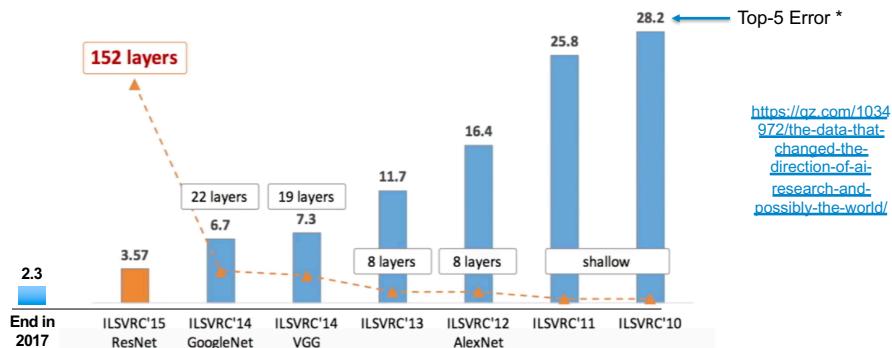
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Some of the Best-Known Datasets																								
<i>Very deep neural networks work best when trained on very large datasets!</i>																								
<ul style="list-style-type: none"> • MNIST: Handwritten digits, 60000 Training Images, 10000 Test Images • CIFAR-10 / CIFAR-100: 50k Training, 10k Test Images of 10 (CIFAR-10) or 100 (CIFAR-100) classes 																								
Color Images are 32x32, Task: Classification https://www.cs.toronto.edu/~kriz/cifar.html <table border="1" style="margin-left: auto; margin-right: auto;"> <tr> <td>airplane</td> <td></td> </tr> <tr> <td>automobile</td> <td></td> </tr> <tr> <td>bird</td> <td></td> </tr> </table>											airplane		automobile		bird									
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<ul style="list-style-type: none"> • Imagenet: > 15 Million Images in 20,000 classes! https://en.wikipedia.org/wiki/ImageNet <table border="1" style="margin-left: auto; margin-right: auto;"> <tr> <td></td> <td>mammal</td> <td>→</td> <td>placental</td> <td>→</td> <td>carnivore</td> <td>→</td> <td>canine</td> <td>→</td> <td>dog</td> <td>→</td> <td>working dog</td> <td>→</td> <td>husky</td> </tr> </table>												mammal	→	placental	→	carnivore	→	canine	→	dog	→	working dog	→	husky
	mammal	→	placental	→	carnivore	→	canine	→	dog	→	working dog	→	husky											

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The Image Large-Scale Visual Recognition Challenge

From 2010 to 2017, ImageNet has been running an annual competition (ILSVRC: Image Large Scale Visual Recognition Challenge) in visual recognition where participants are presented with the challenge of classifying 1.2 million images belonging to 1000 different classes from Imagenet dataset.



* The Top-5 error is the percentage of the time that the classifier did not include the correct class among its top 5 guesses.

<https://www.youtube.com/watch?v=DAOcjcFr1Y>

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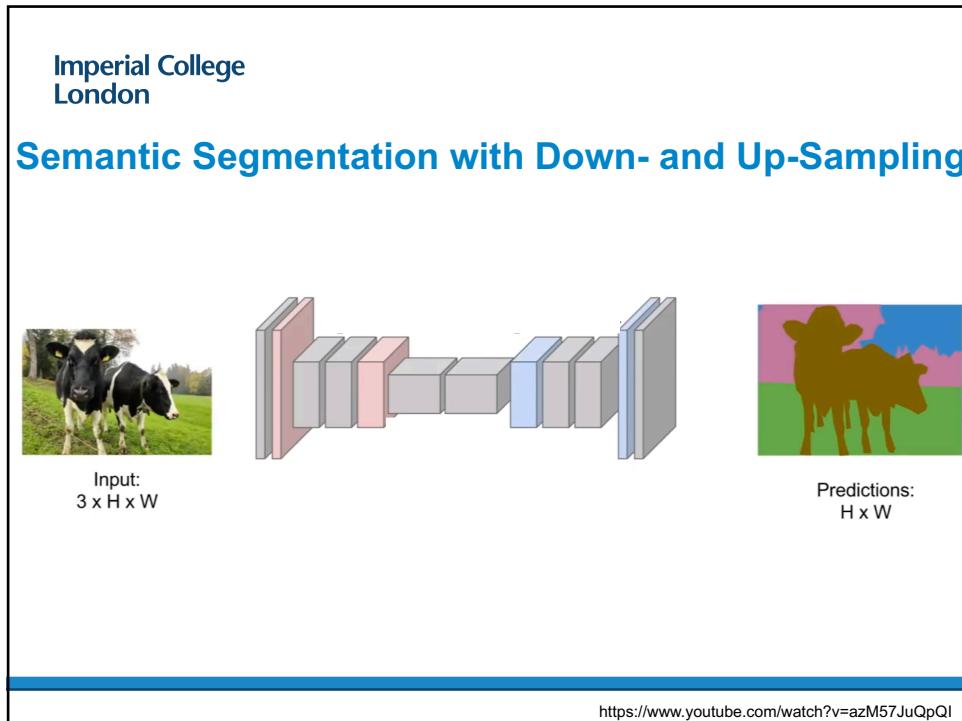
Transfer Learning (1)

The most well-known CNN designs are available on-line and have been successfully trained on very large number of images (1,000,000s).

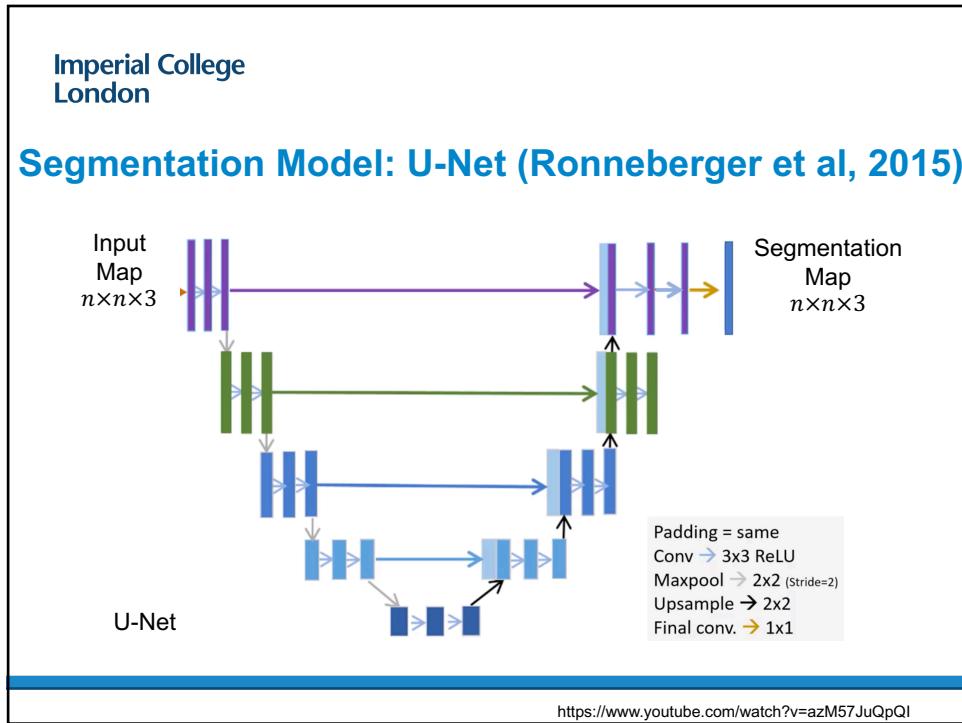
In many applications we often work from a relatively small number of images.

Why not start from an existing trained CNN sharing an objective of a similar nature (such as Classification) and tailor it to our application?

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