

Reminder CNNs

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London

Computer Vision Topics



Image Segmentation

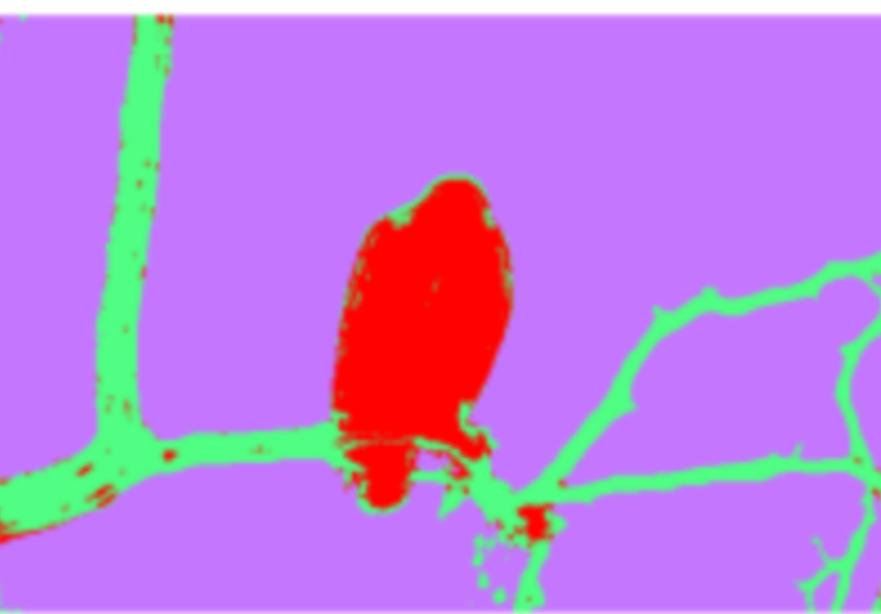
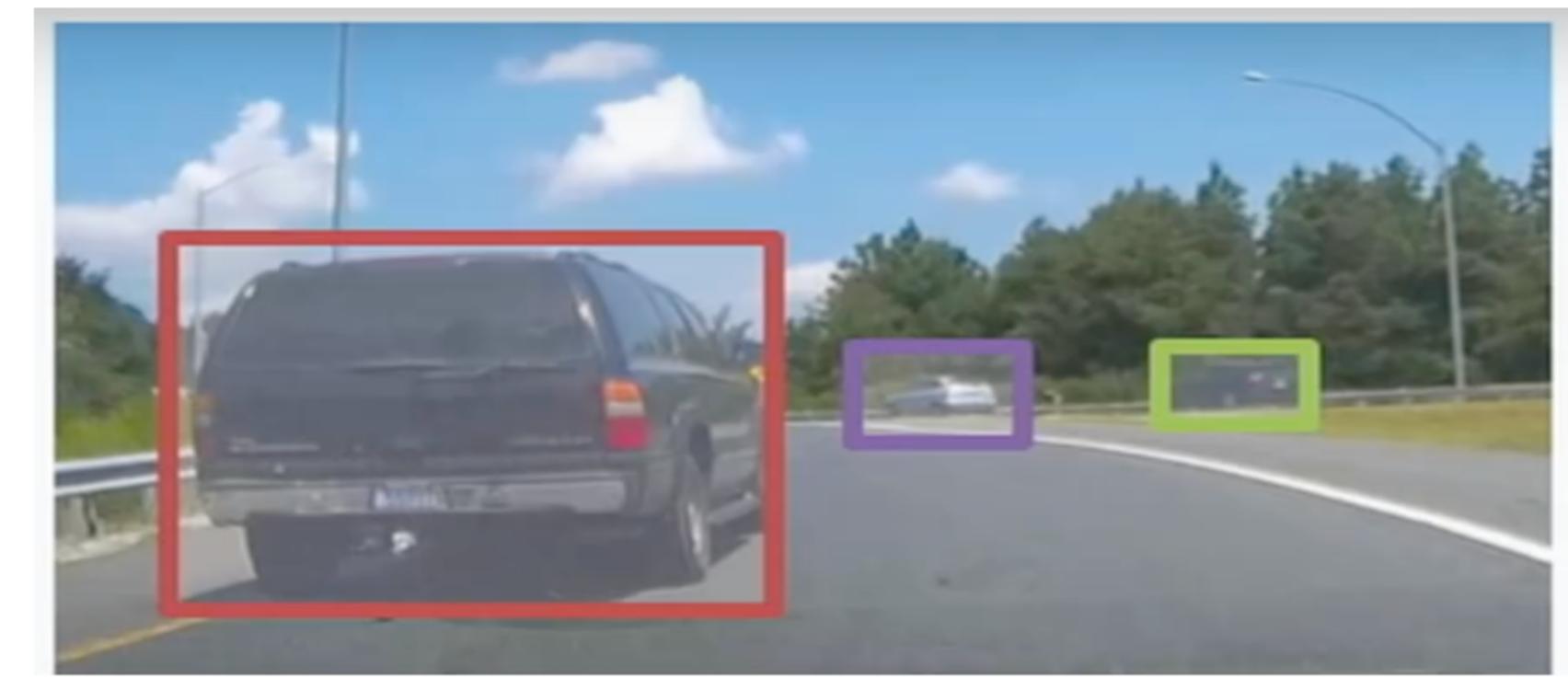


Image Classification



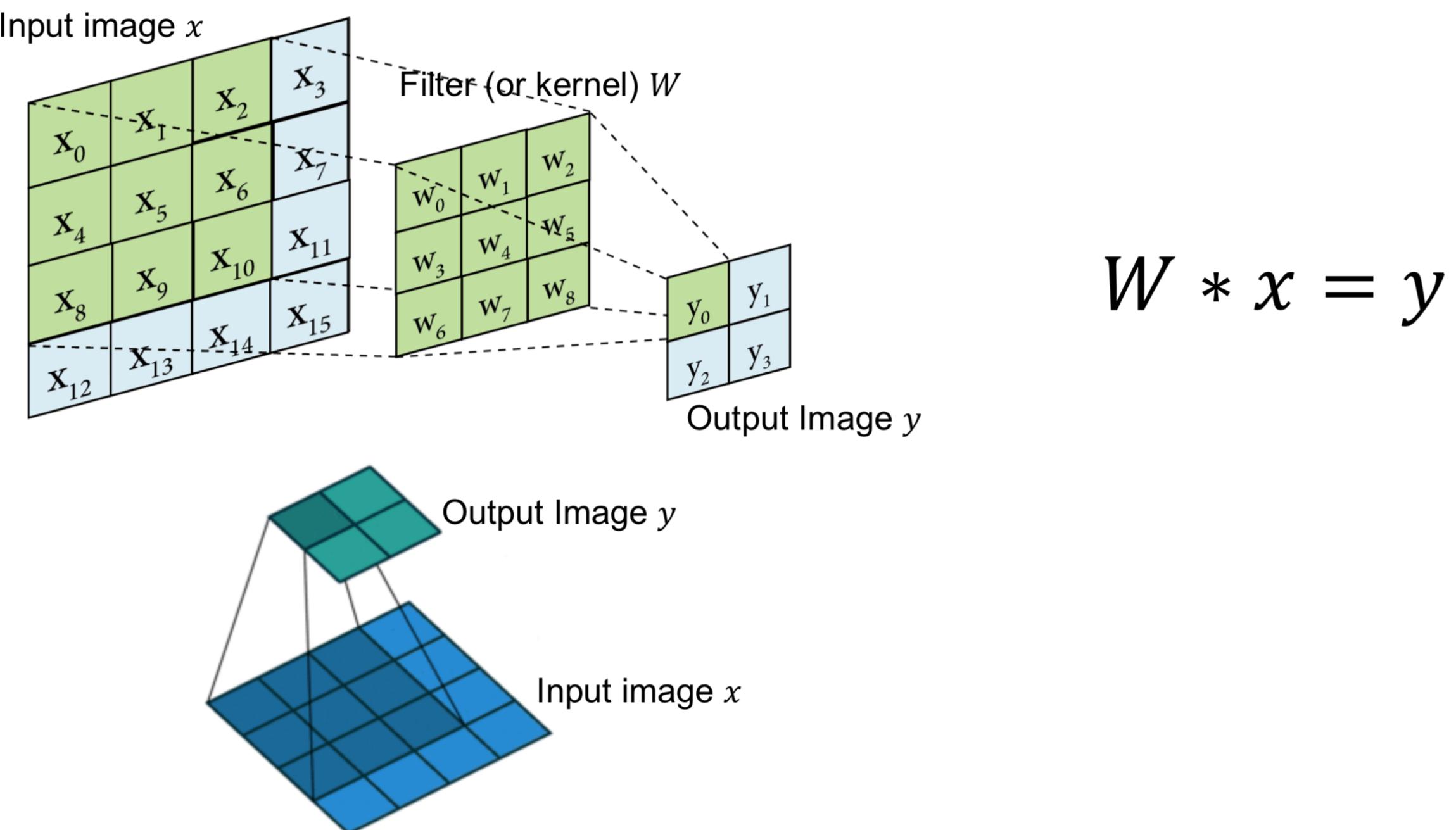
Object Localization (1 object) or Detection (several objects)

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Idea Behind Convolutional Neural Networks:

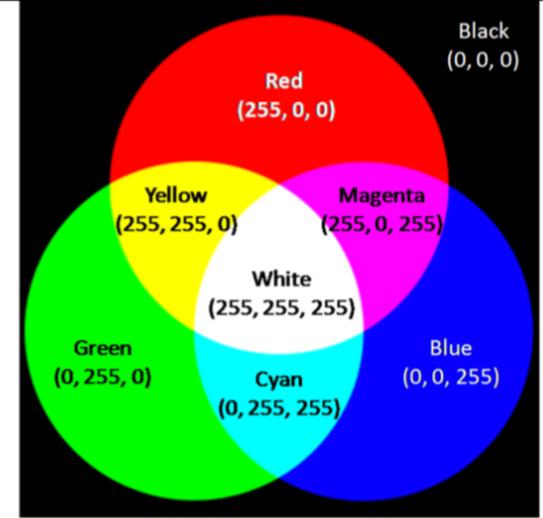
Parametrize the Filter and Optimize its Coefficients Based on the Objective!



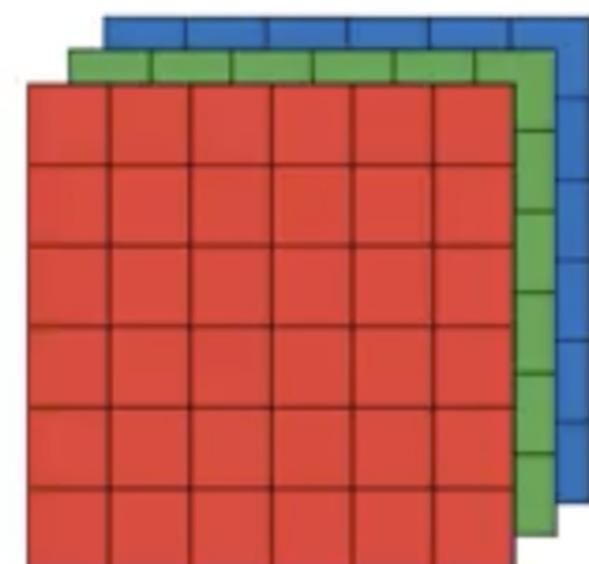
$$y_0 = w_0x_0 + w_1x_1 + w_2x_2 + w_3x_4 + w_4x_5 + w_5x_6 + w_6x_8 + w_7x_9 + w_8x_{10}$$

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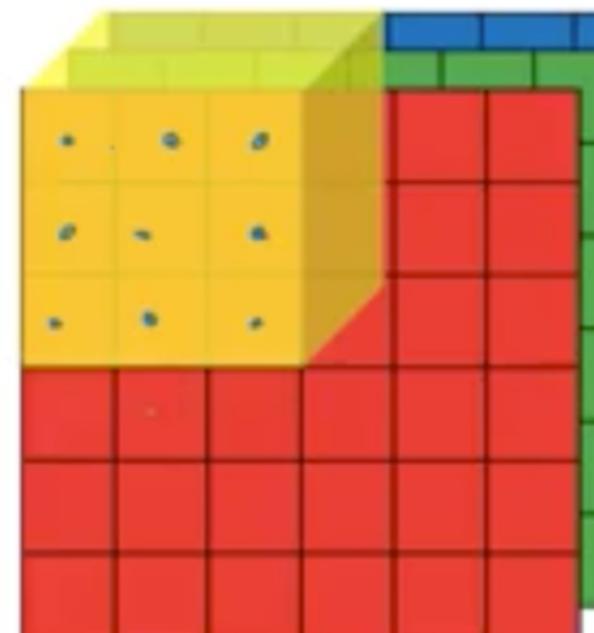
If the Two-Dimensional image has 3 RGB channels...



6x6x3

$$\begin{matrix} * & \begin{matrix} 3 \times 3 \times 3 \end{matrix} & = & \begin{matrix} 4 \times 4 \times 1 \end{matrix} \end{matrix}$$

27 operations
each time
(28 if bias term)

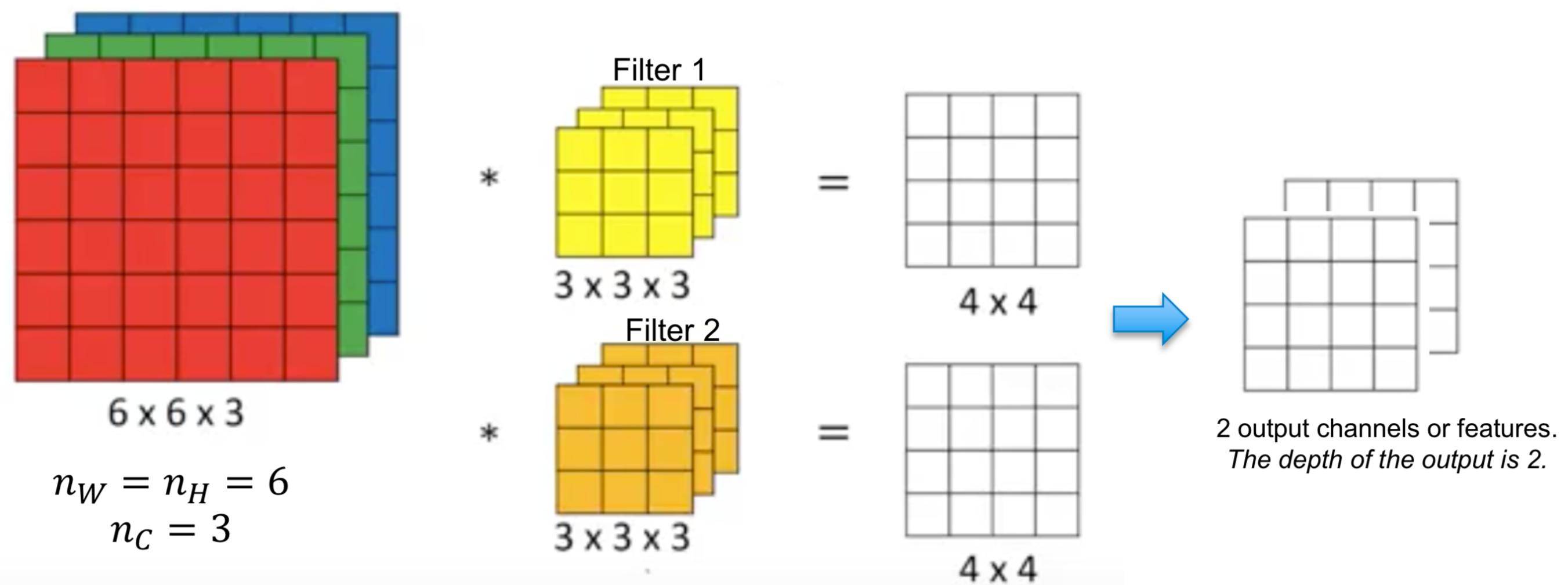


Each slice of the filter can detect different features in each of the 3 colour channels, for instance vertical edges in Red, horizontal edges in Green and vertical edges in Blue!

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If Different Filters are Applied ...



Exercise: if my input Image is 50x50 with 10 channels, and if I apply 25 filters each of size 3x3x10 with a stride of 1, no padding, what is the size and number of channels of the output image?

Answer: size is 48x48. with 25 channels

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Example of Max Pooling Layer (for Downsampling)

1	3	2	1
2	9	1	1
1	3	2	3
5	6	1	2

4 x 4



9	2
6	3

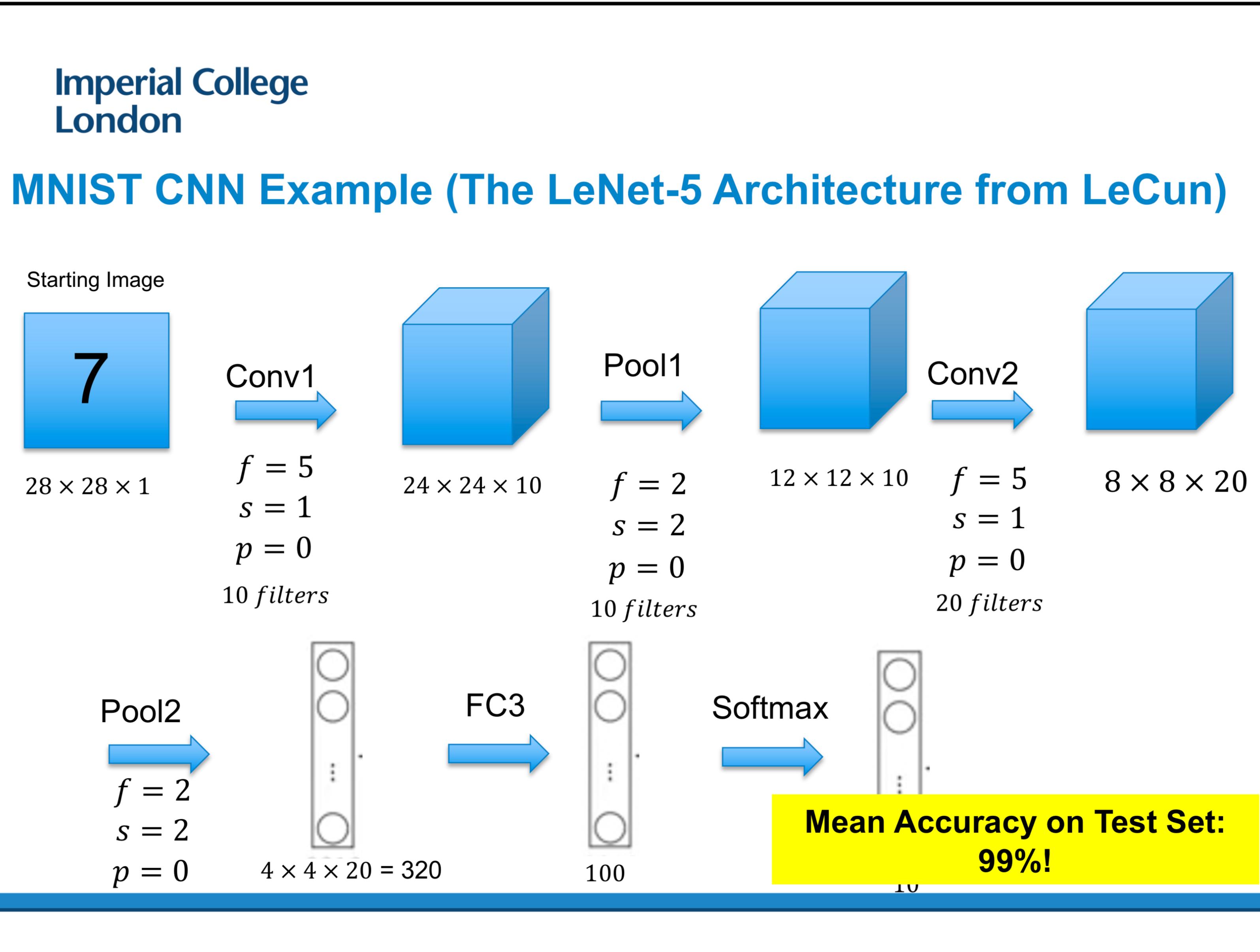
2 x 2

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

Stride s and
Filter size f are
the main
hyperparameters
for pooling.

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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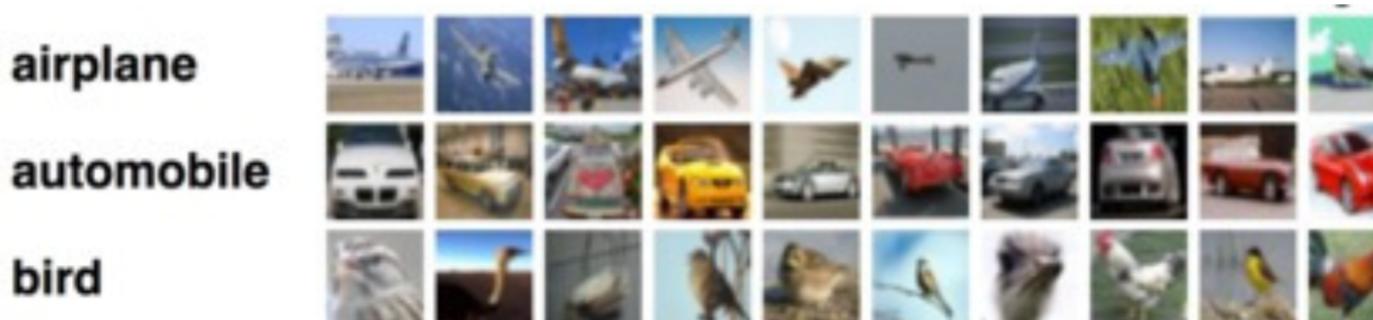
Some of the Best-Known Datasets

Available on
`torchvision.datasets`

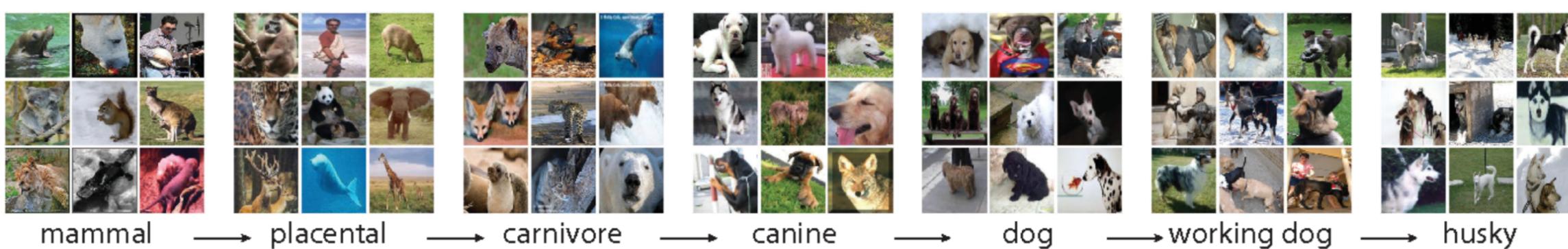
Very deep neural networks work best when trained on very large datasets!

- **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>



- **Imagenet:** > 15 Million Images in 20,000 classes! <https://en.wikipedia.org/wiki/ImageNet>

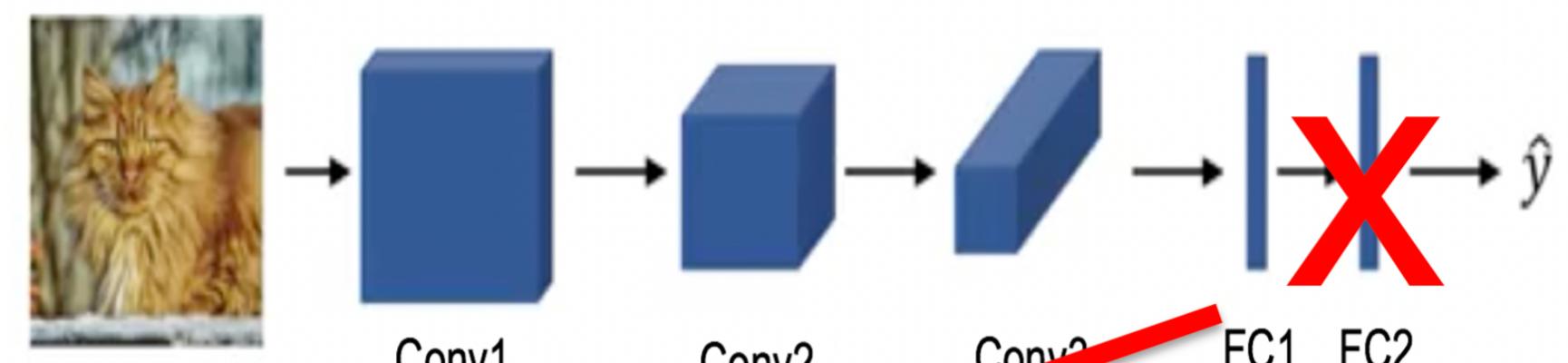


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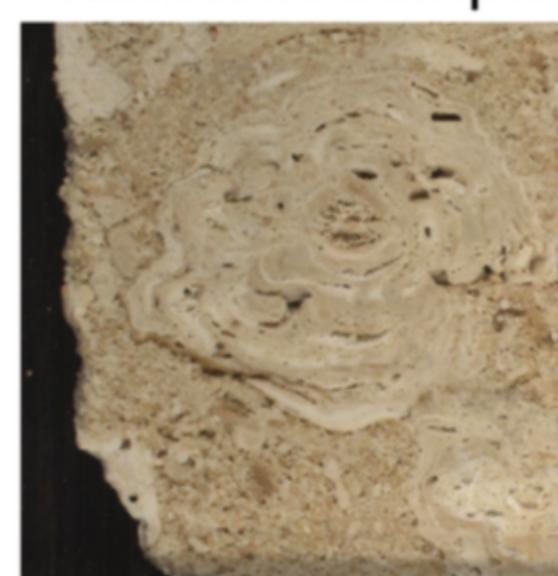
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Example of Transfer Learning Approach

Take existing trained network such as Inception v3 model, trained on ImageNet dataset for differentiating between 1,000 different classes of images.



Re-train last layer of the network on images of interest, such as pictures of carbonate cores .



From Sharinia Kanagandran and Cedric John, Imperial College

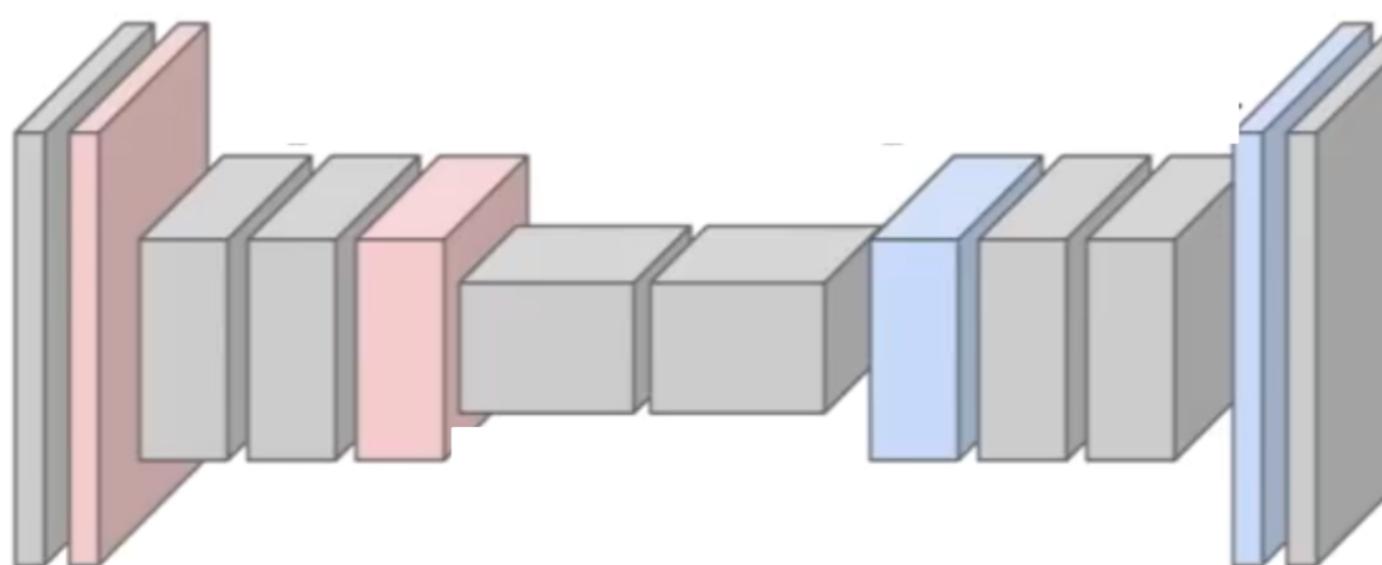
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Semantic Segmentation with Down- and Up-Sampling



Input:
 $3 \times H \times W$



Predictions:
 $H \times W$

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Segmentation Model: U-Net (Ronneberger et al, 2015)

