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# Applied Machine Learning with Big Data

## “EE 6973”



Topic:  
Convolution Neural Networks

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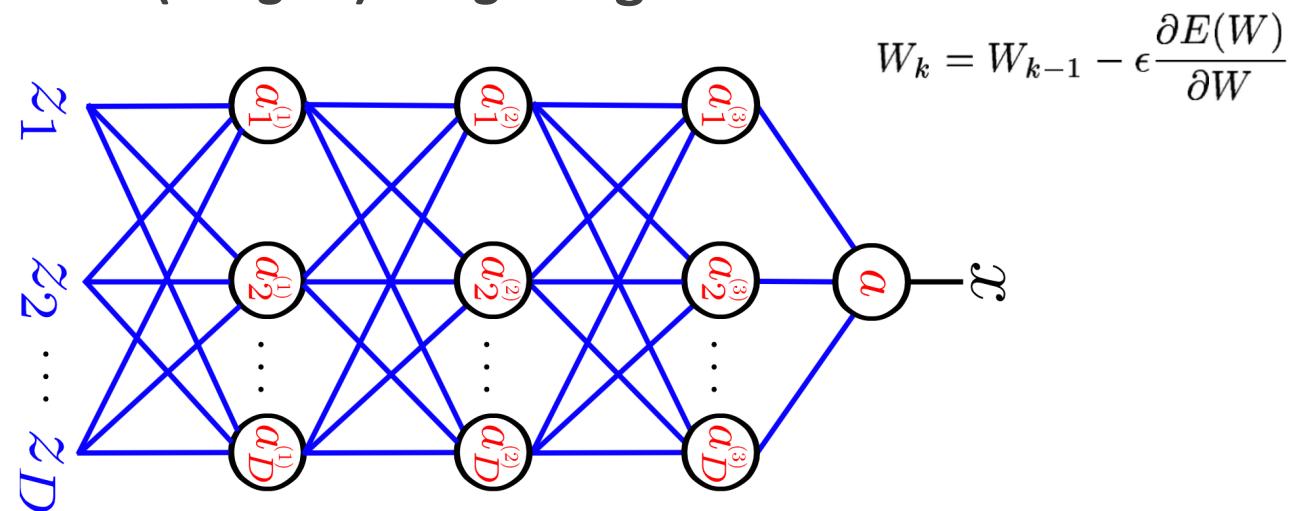
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# Stochastic Gradient Descent (SGD)

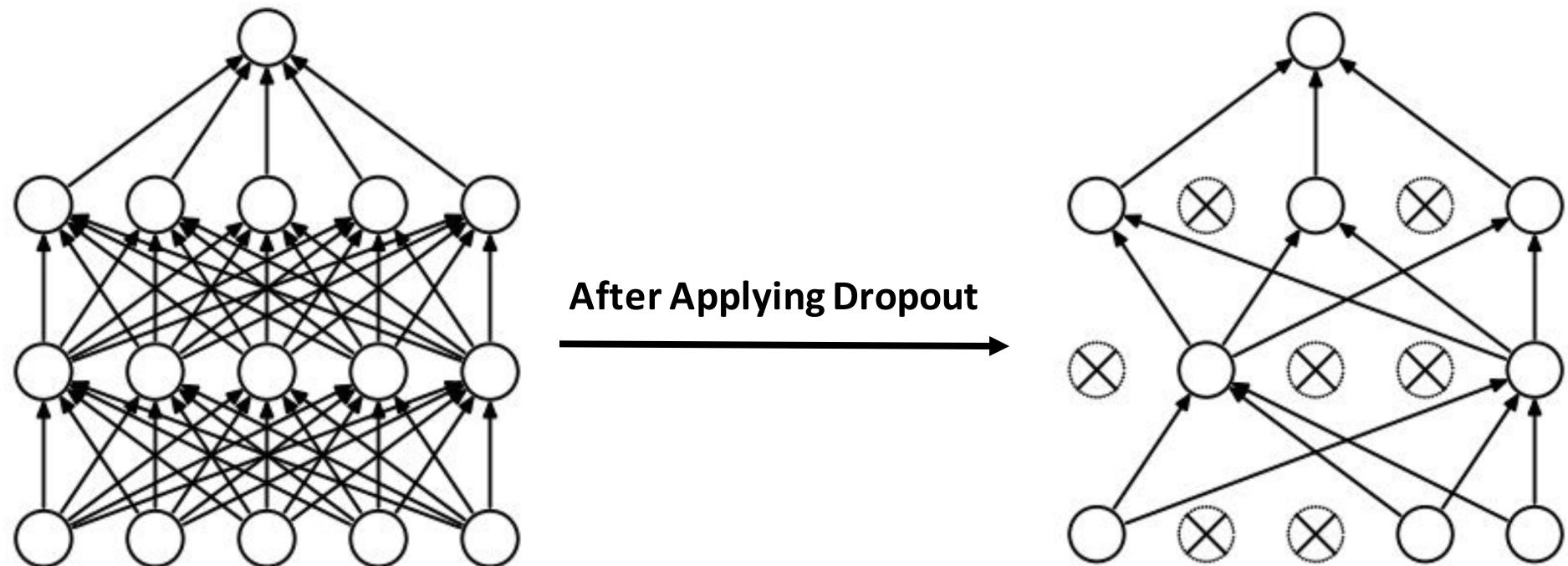
Loop:

1. Sample a batch of data with their labels
2. Forward Propagate it through the graph, calculate the error
3. Backpropagate to calculate the gradients
4. Update the parameters (weights) using the gradient



# Dropout

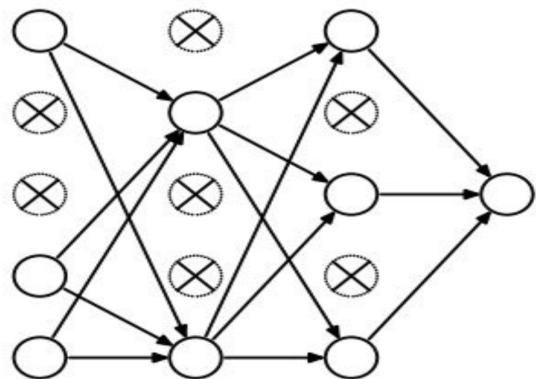
**Randomly set some neurons to zero in the forward pass**



Srivastava et al., 2014, <https://www.cs.toronto.edu/~hinton/absps/JMLRdropout.pdf>

# Dropout

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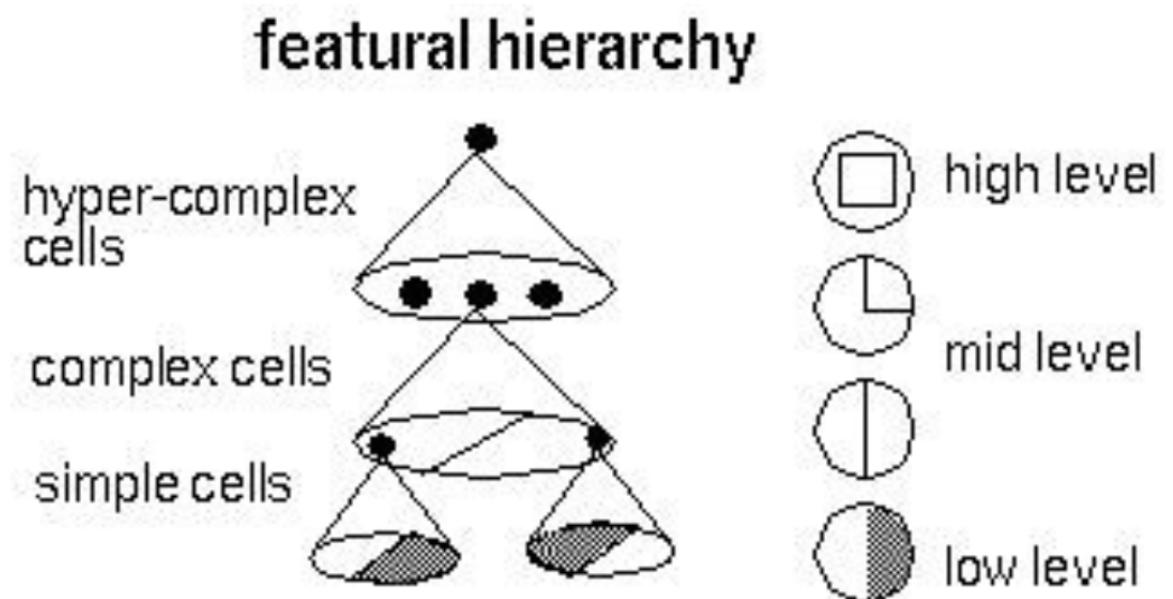
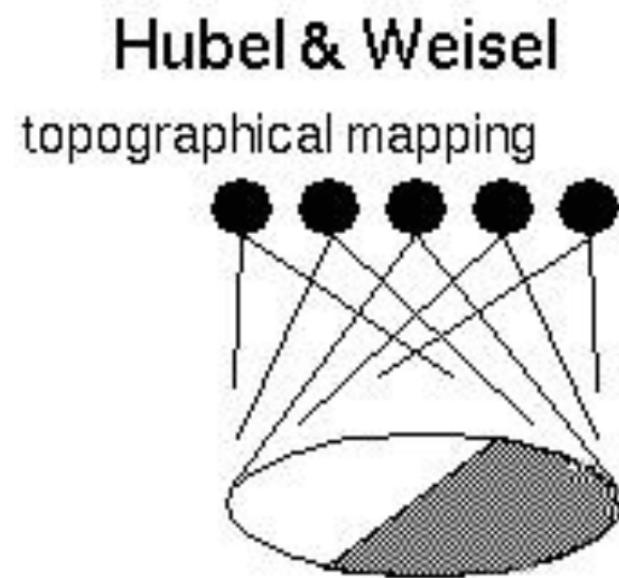


**Forces the network to have  
a redundant representation**



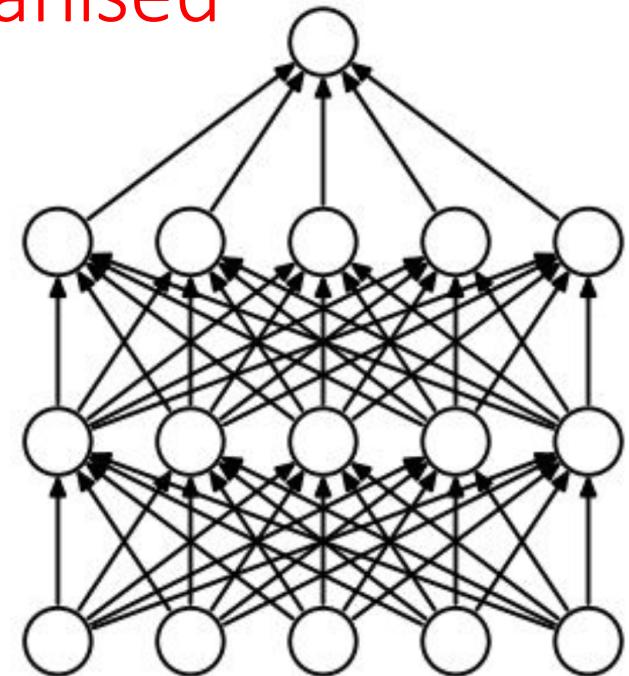
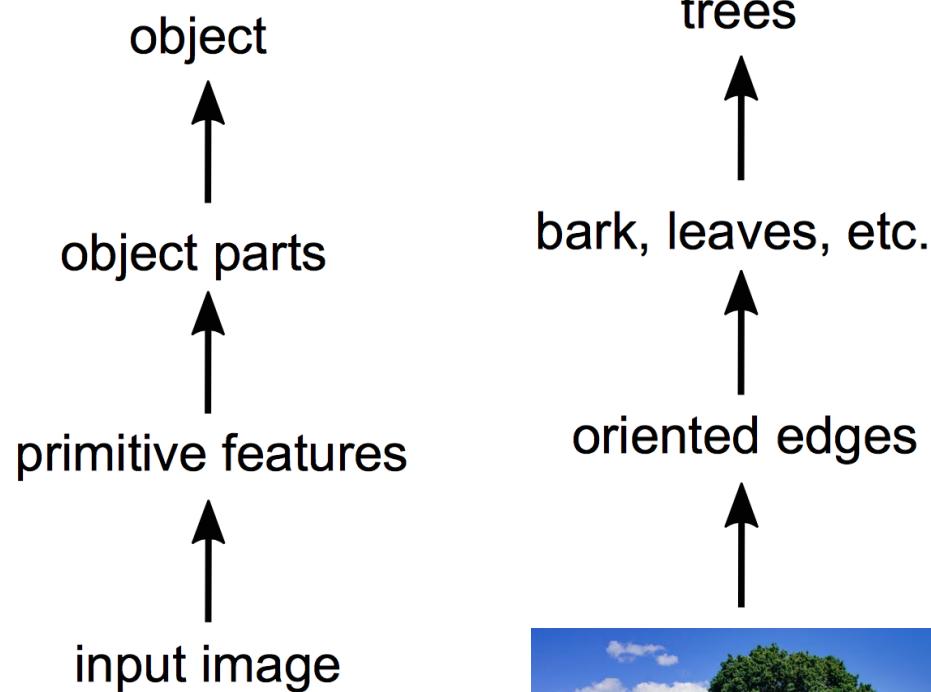
# Vision: Hierarchical Organization (Year: 1962)

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# Why use hierarchical multi-layered models?

## Biological vision is hierachically organised



# What's wrong with standard neural networks?

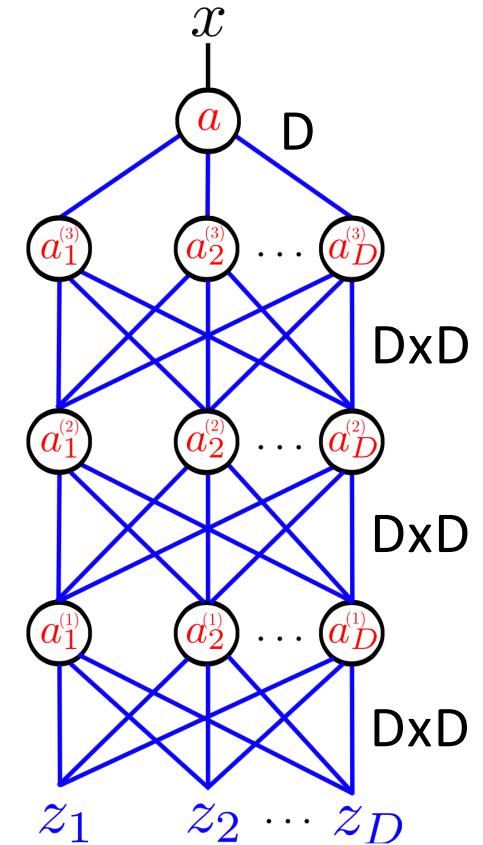
## Hard to Train

How many parameters does this network have?

Number of Parameters =  $3 \times (D \times D) + D$

For a small  $D = 32 \times 32 = 1024$  MNIST image:

Number of Parameters =  $3 \times (1024 \times 1024) + 1024$   
 $\sim 3 \times 10^6$



# Architecture of LeNet-5, Convolution Neural Network

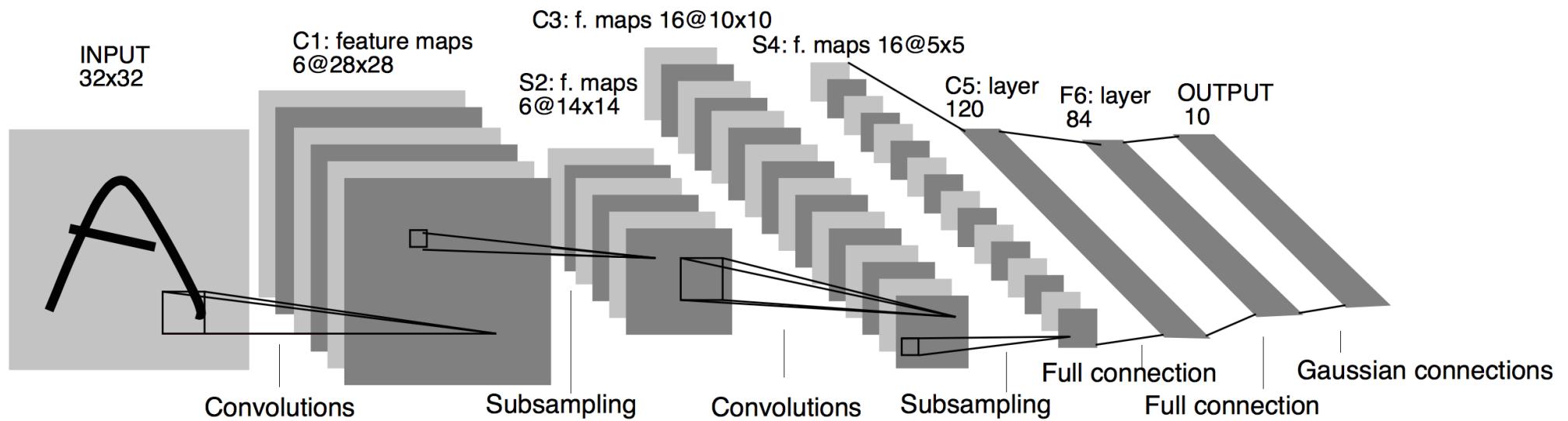
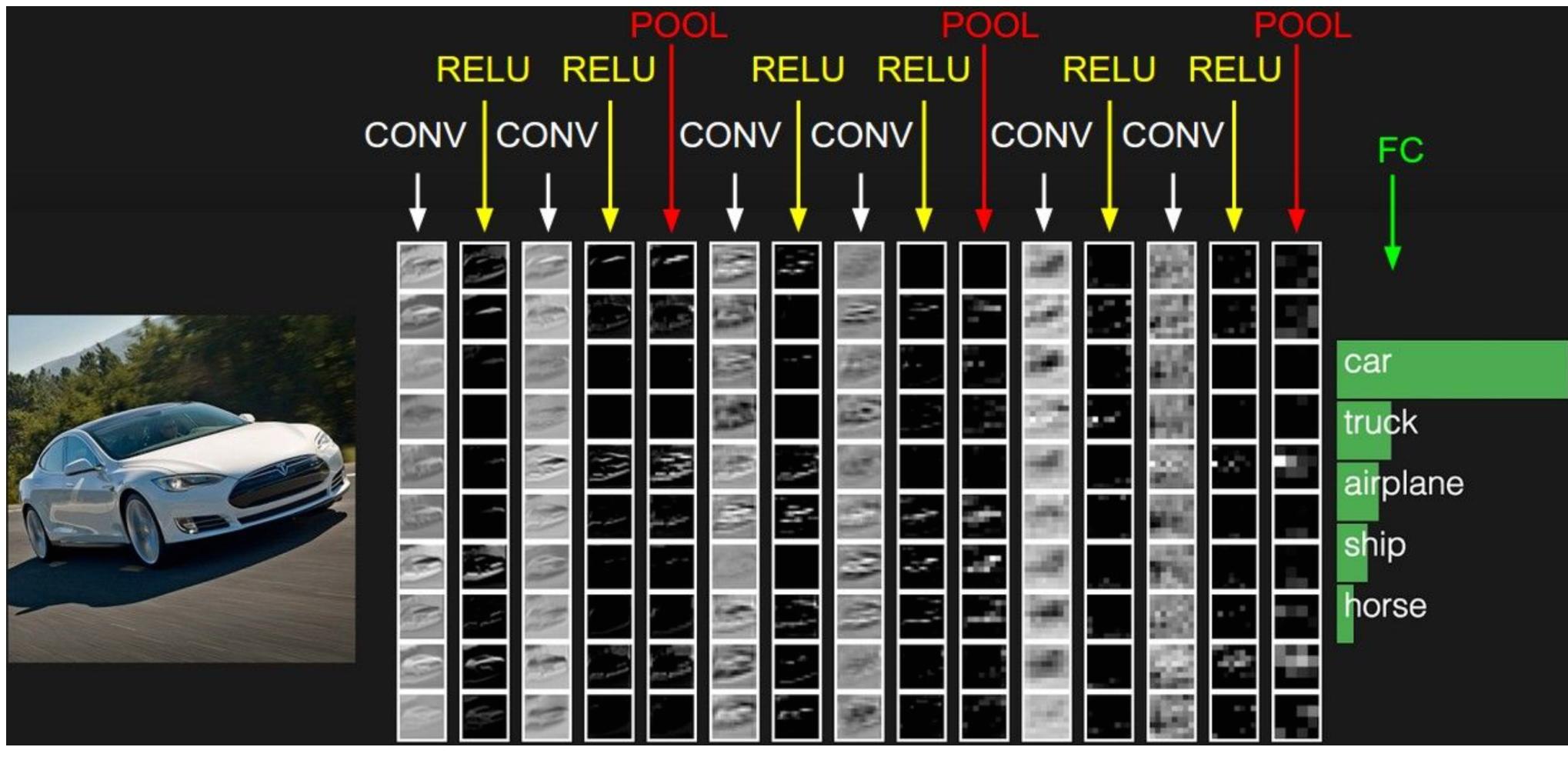


Fig. 2. Architecture of LeNet-5, a Convolutional Neural Network, here for digits recognition. Each plane is a feature map, i.e. a set of units whose weights are constrained to be identical.

Proc. Of the IEEE, November 1998, "Gradient-Based Learning Applied to Document Recognition"



# Review: What is convolution?

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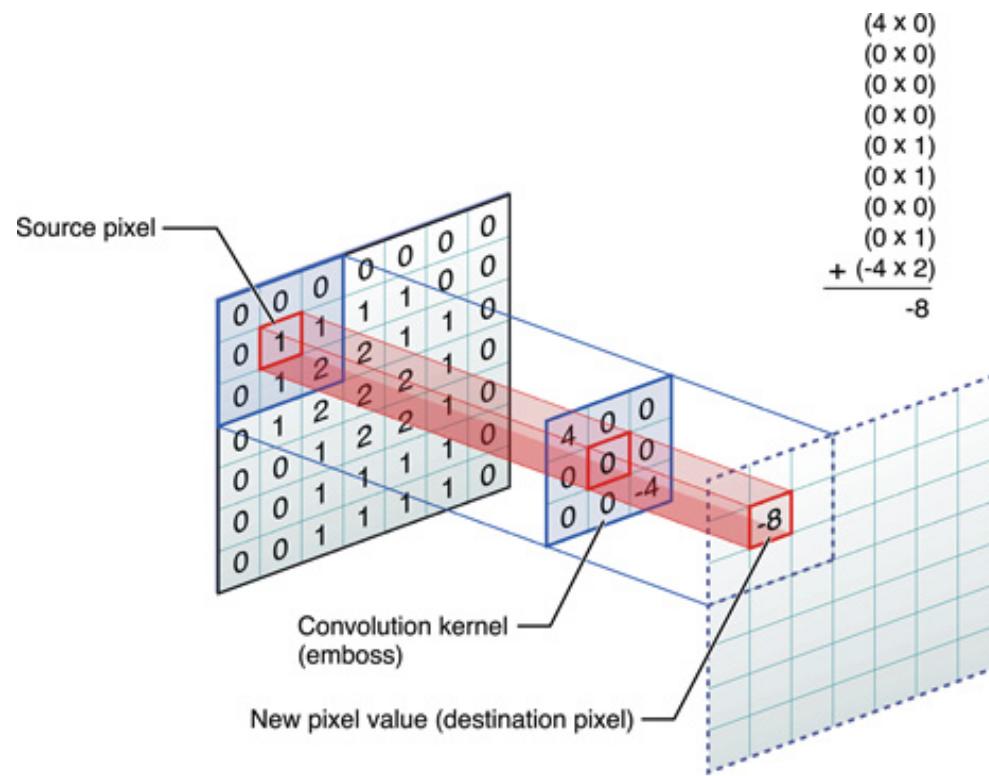
- Convolution is an important operation from signal processing
- A convolution is an integral that expresses the amount of overlap of one function as it is shifted over another function .

$$f * g = \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau = \int_{-\infty}^{\infty} g(\tau) f(t - \tau) d\tau$$

- 2 Dimensional Discrete Function (Image)

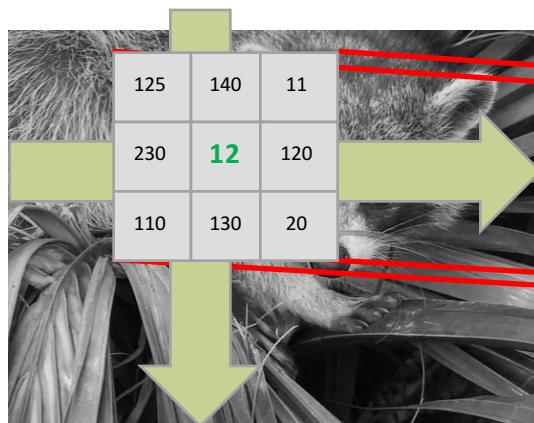
$$f[x, y] * g[x, y] = \sum_{n_1=-\infty}^{\infty} \sum_{n_2=-\infty}^{\infty} f[n_1, n_2] \cdot g[x - n_1, y - n_2]$$

# 2-Dimensional Convolution

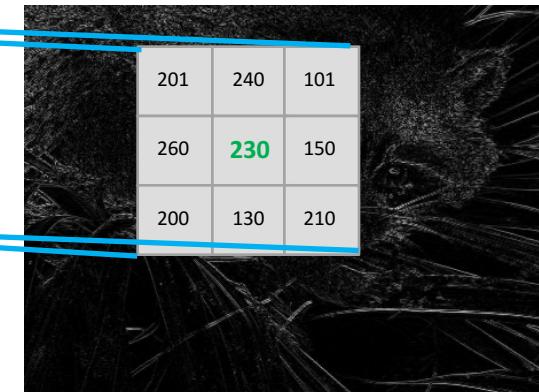


# Example: 2-Dimensional Convolution

A convolution is an integral (**discrete signals :Matrix Dot Product**) that expresses the amount of overlap of one function as it is shifted over another function



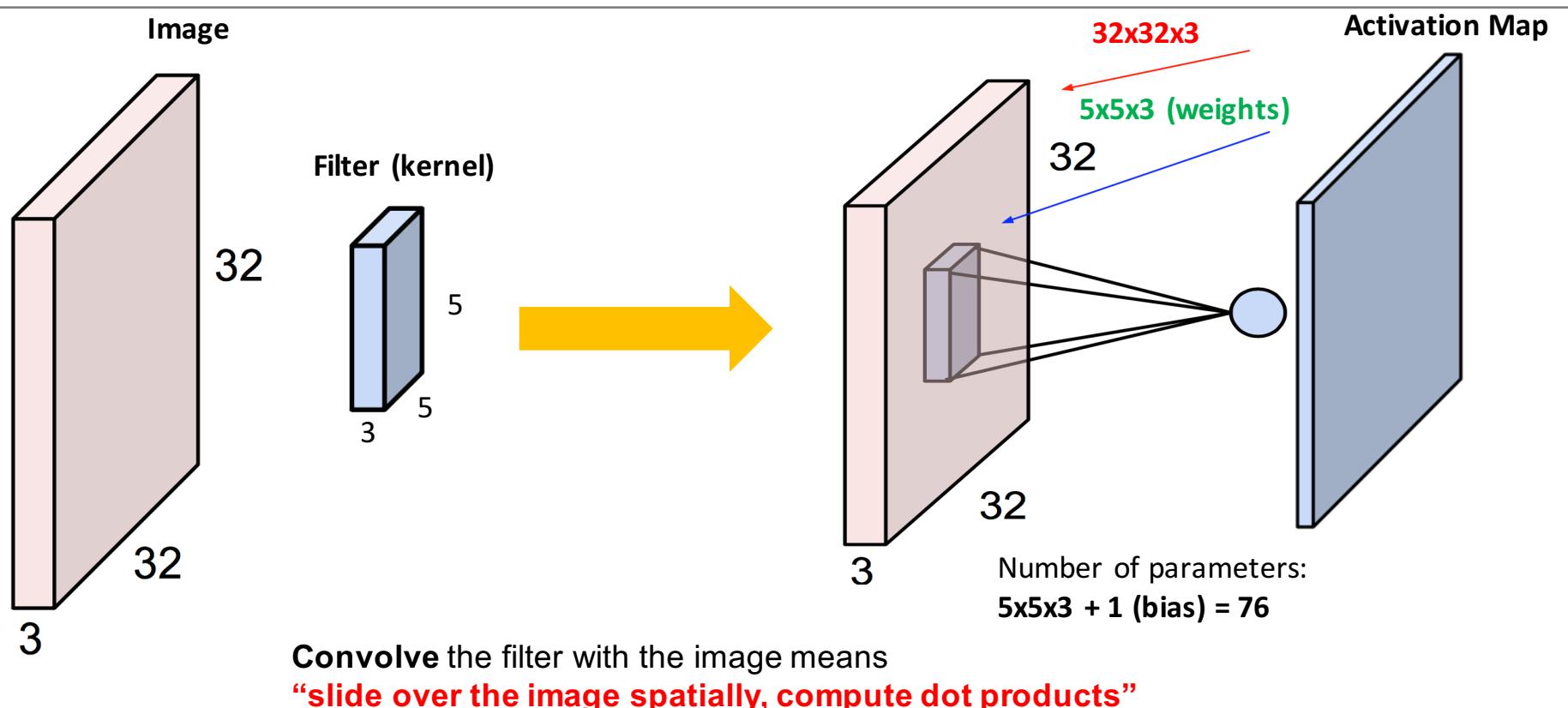
-1	-2	-1
0	0	0
1	2	1

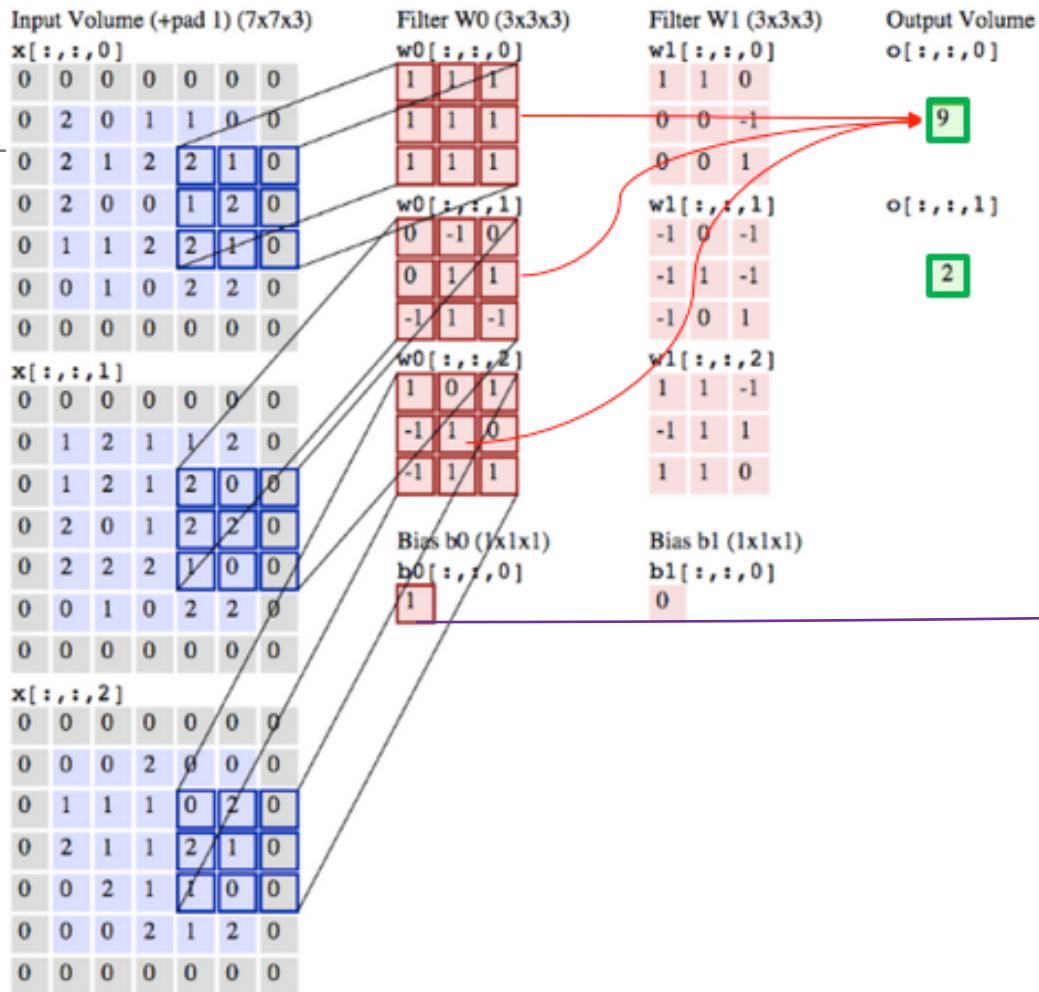


-1	0	1
-2	0	2
-1	0	1



# Convolution Layer



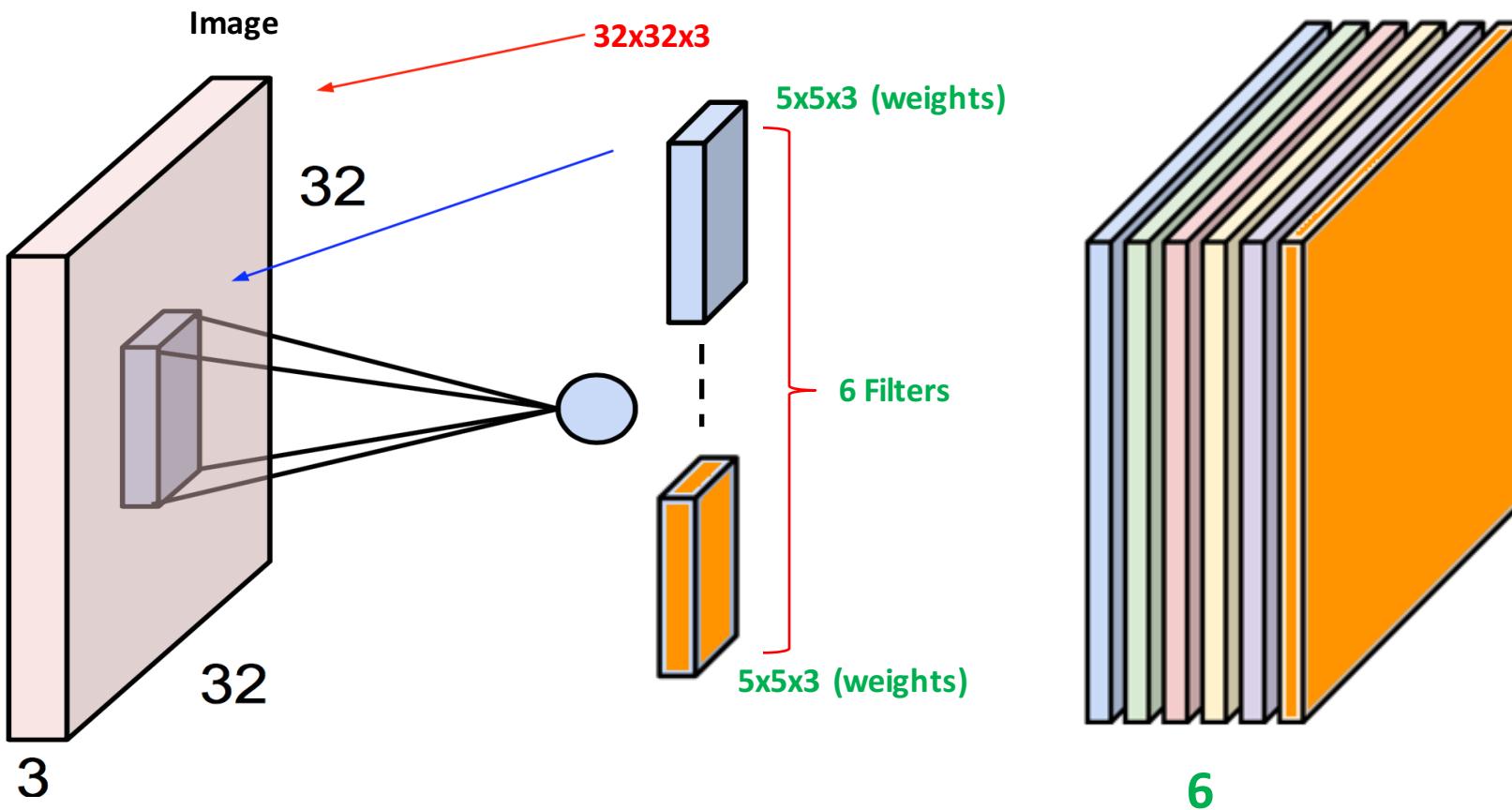


$$(2 \times 1) + (1 \times 1) + 0 + (1 \times 1) + (2 \times 1) + 0 + (2 \times 1) + (1 \times 1) + 0 = 9$$

$$0 + 0 + 0 + (2 \times 1) + 0 + (1 \times -1) + 0 + 0 = 1$$

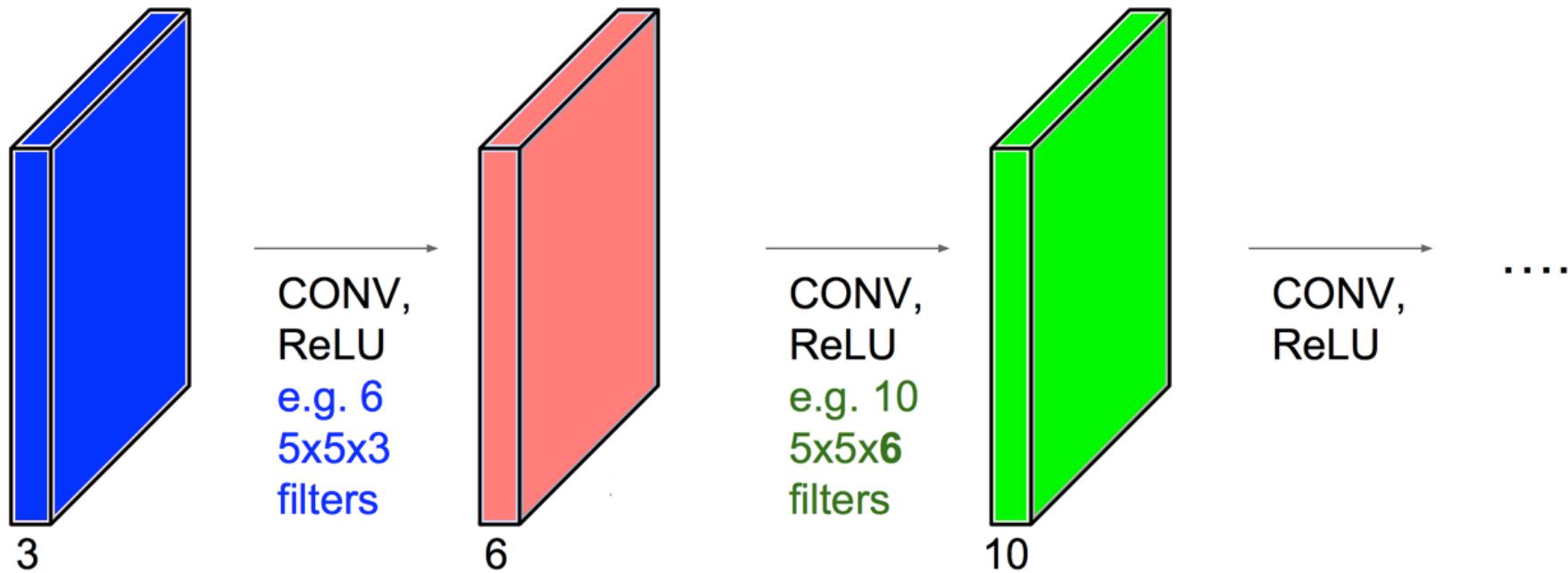
$$0 + 0 + 0 + (2 \times -1) + (1 \times 1) + 0 + (1 \times -1) + 0 + 0 = -2$$

$$1 + 1 + (-2) + 9 = 9$$

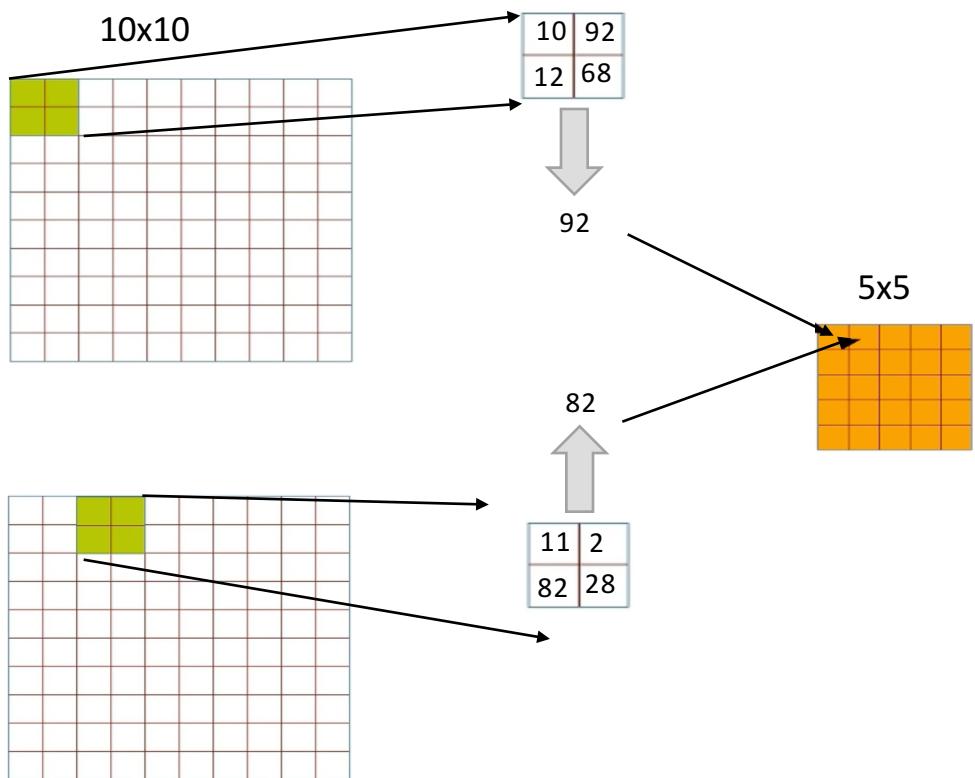


# Convolution Network

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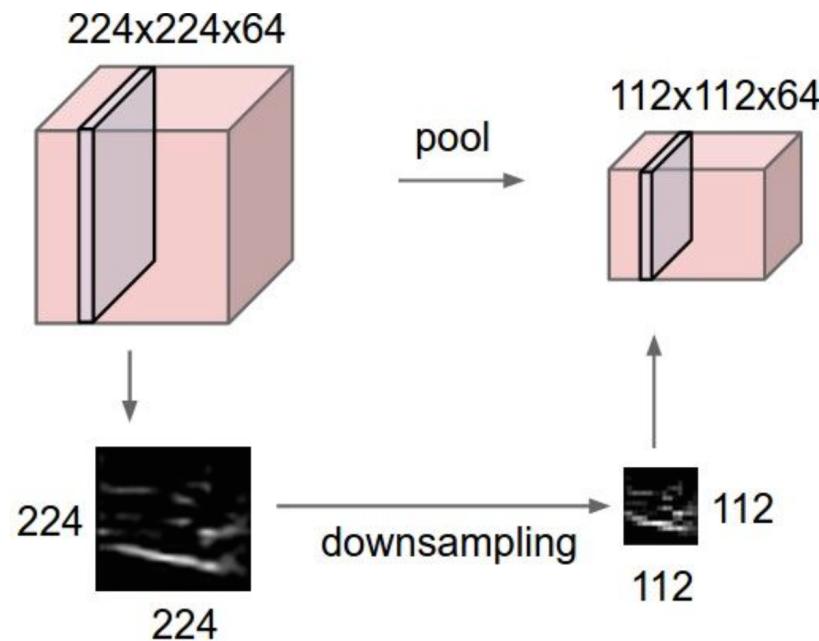
# Downsampling= Max Pooling

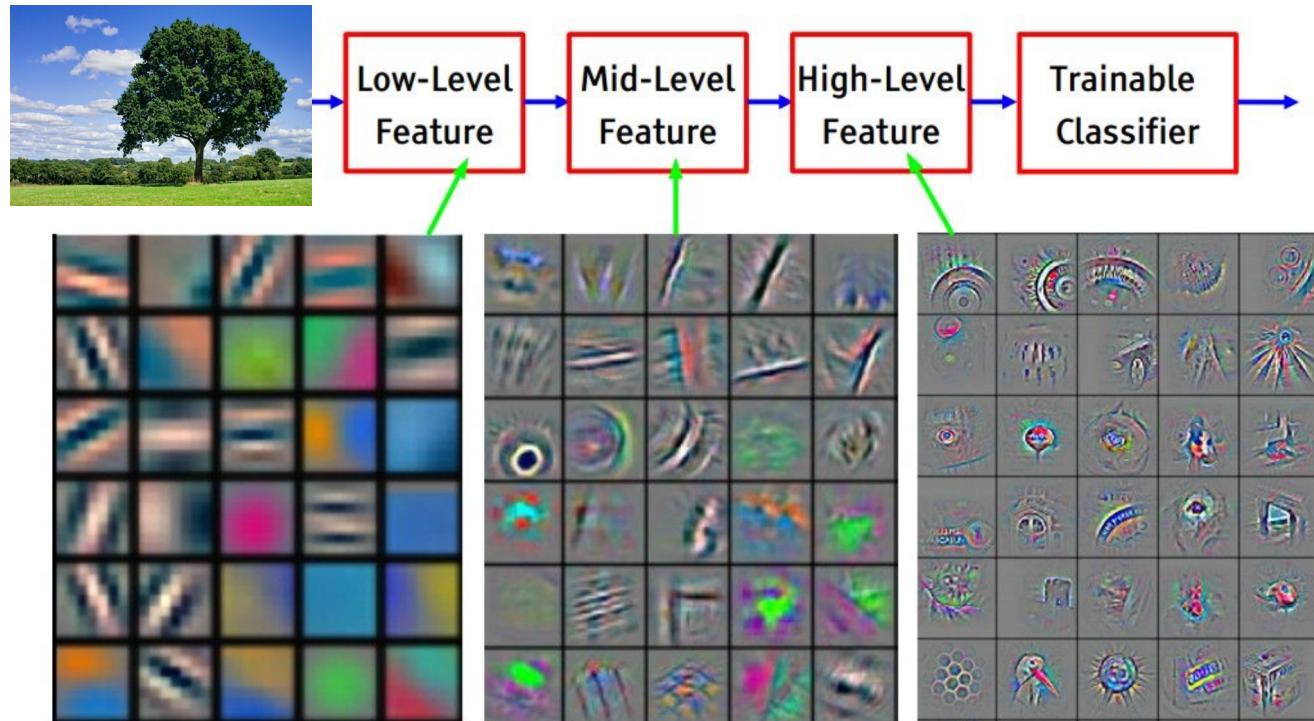


**“Max Pooling” Layers to extract the  
“best” local feature**

# Pooling Layer

- Makes the representations smaller and more manageable
- Operates over each activation map independently





Feature visualization of convolution net trained on ImageNet from [Zeiler & Fergus 2013]

