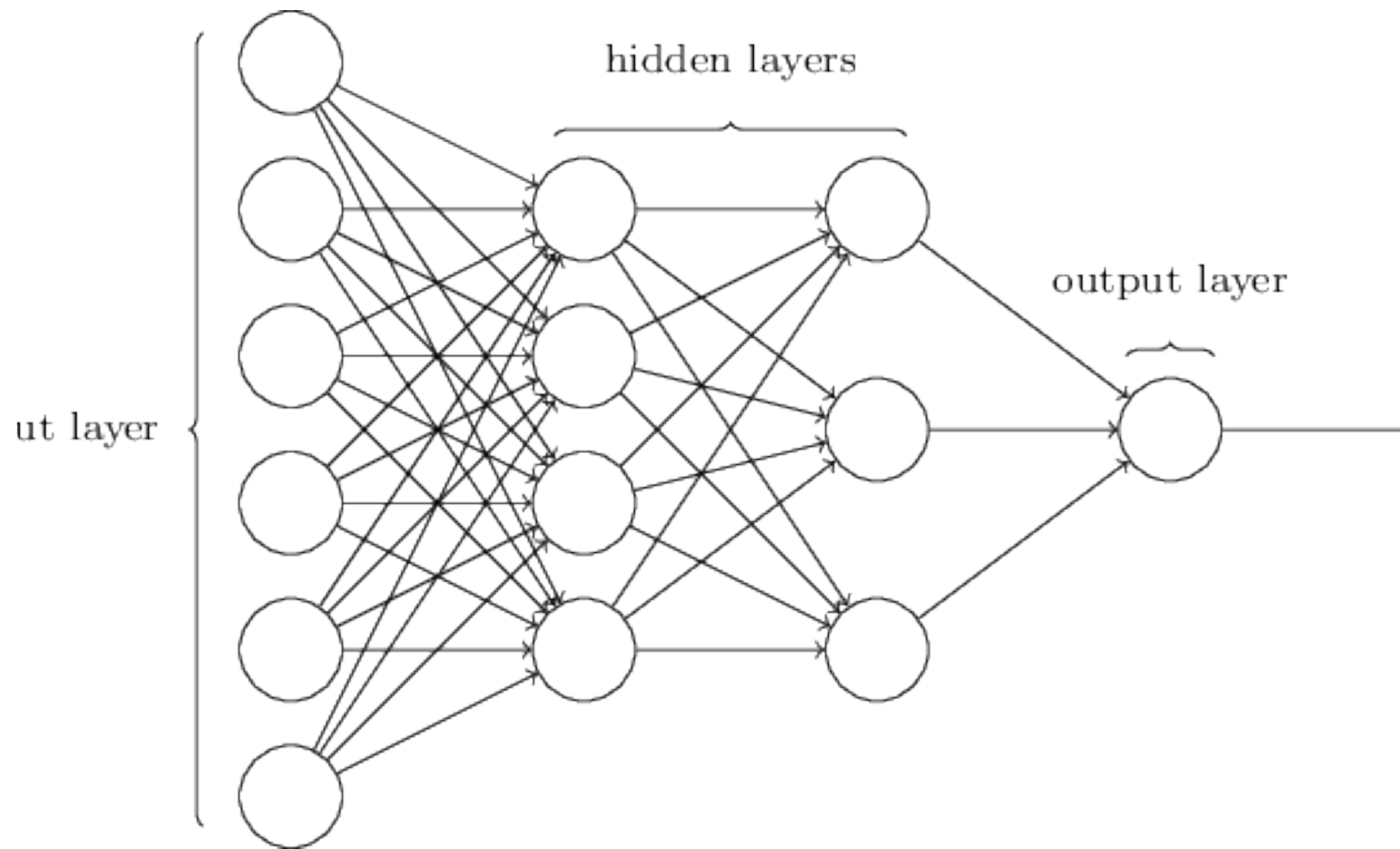


Convolutional Neural Networks

Deep Learning Course
By Felix Kreuk



Recall DNN

(AKA "Fully-Connected")

How Did We Solve MNIST?

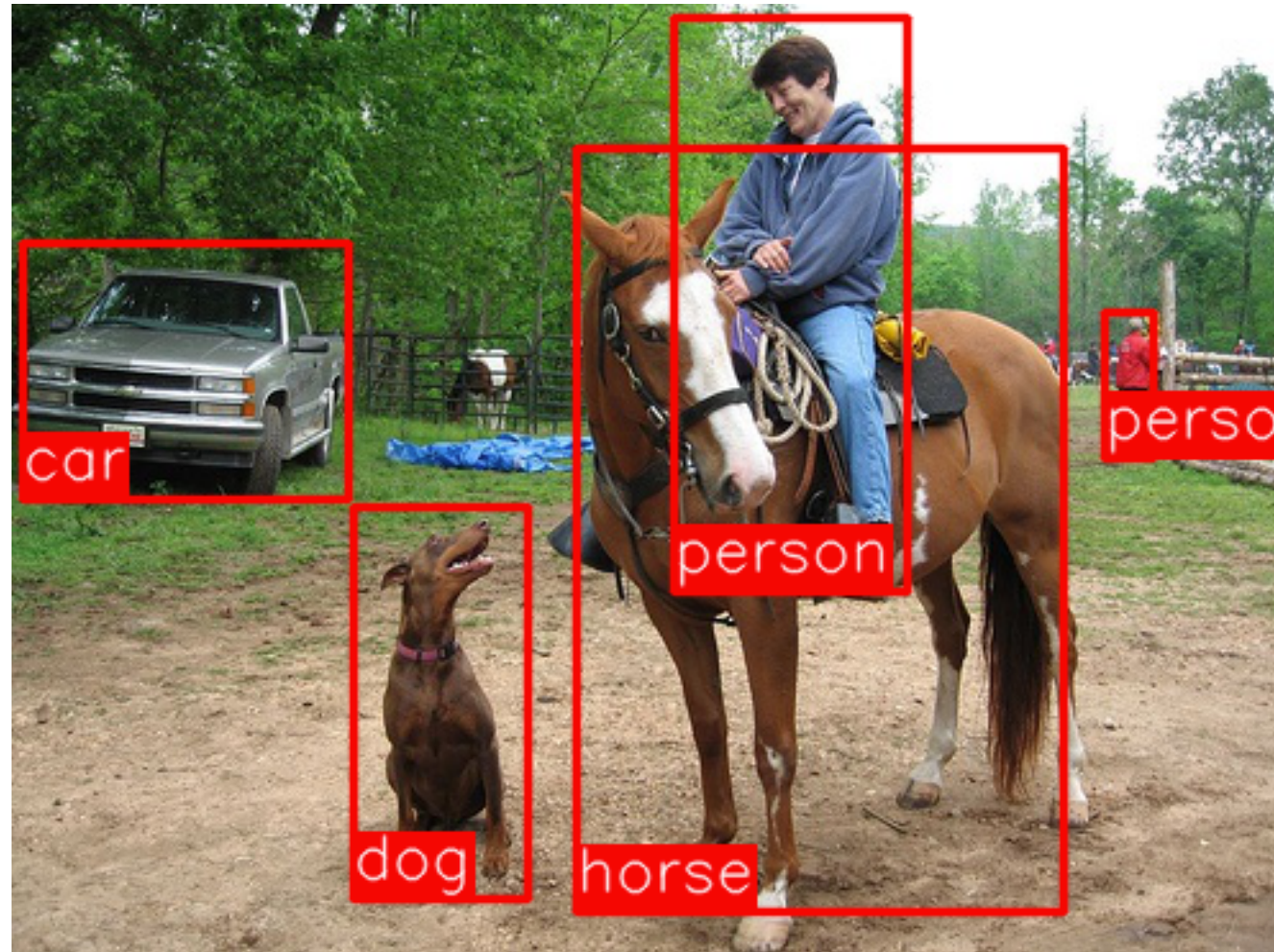
- Transform a 28x28 image to a 1x784 row vector.
- Feed to a DNN.
- Accuracy ~95%.
- Can we do better?

How Did We Solve MNIST?

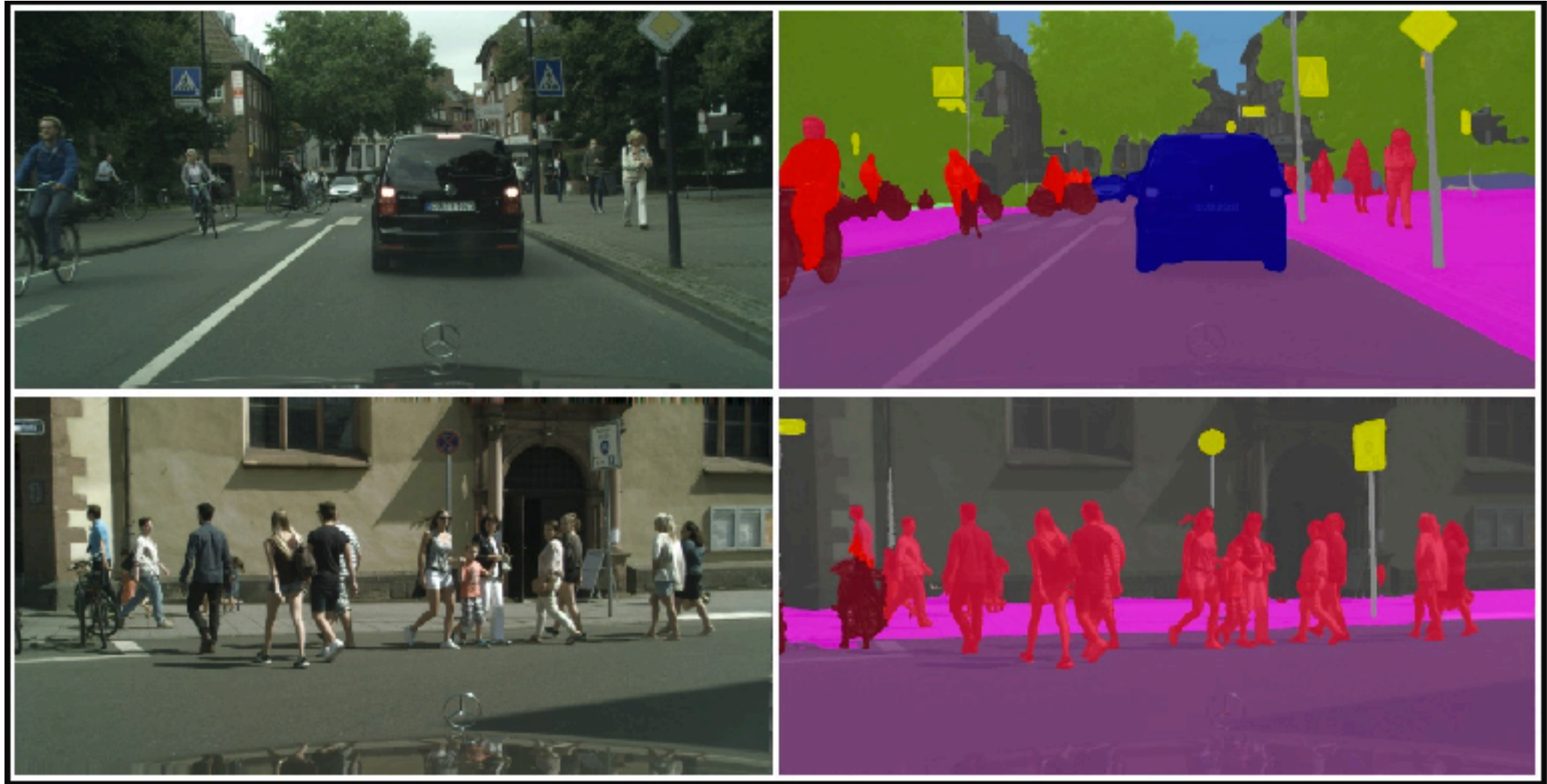
- Transform a 28x28 image to a 1x784 row vector - **what about spatial relationship?**
- Feed to a DNN - **how many parameters? What features did the network learn?**
- Accuracy ~95% - **is this the best we can achieve?**
- Can we do better - **yes.**
- **These are some problems to think about for the next slides.**

Introducing CNNs

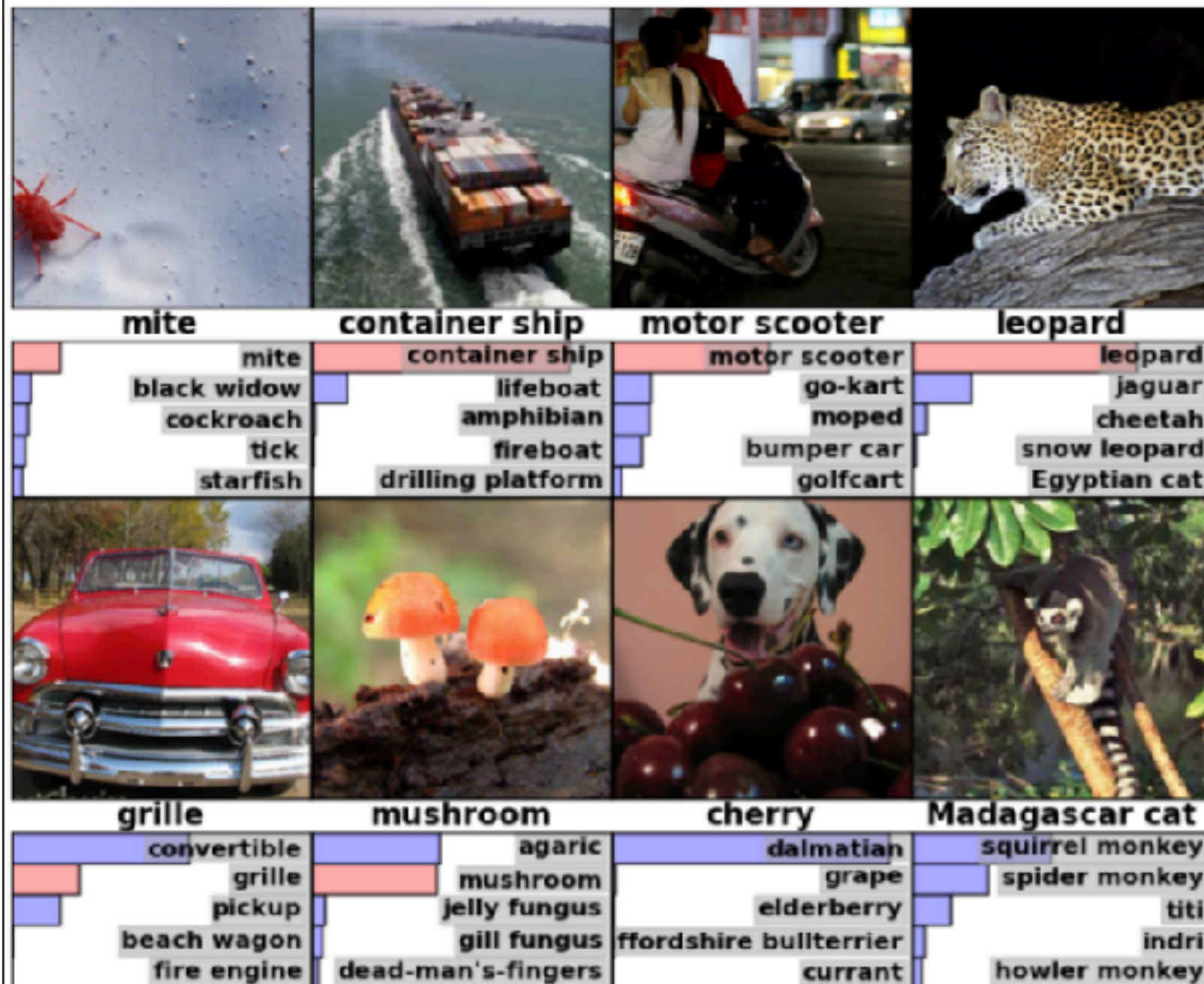
They Are Everywhere



They Are Everywhere



They Are Everywhere



They Are Everywhere

Describes without errors	Describes with minor errors	Somewhat related to the image
 <p data-bbox="455 1126 935 1208">A person riding a motorcycle on a dirt road.</p>	 <p data-bbox="1094 1132 1607 1173">Two dogs play in the grass.</p>	 <p data-bbox="1780 1132 2307 1214">A skateboarder does a trick on a ramp.</p>
 <p data-bbox="460 1678 935 1759">A group of young people playing a game of frisbee.</p>	 <p data-bbox="1081 1671 1668 1753">Two hockey players are fighting over the puck.</p>	 <p data-bbox="1794 1678 2280 1759">A little girl in a pink hat is blowing bubbles.</p>

They Are Everywhere

1 Upload photo

The first picture defines the scene you would like to have painted.



2 Choose style

Choose among predefined styles or upload your own style image.



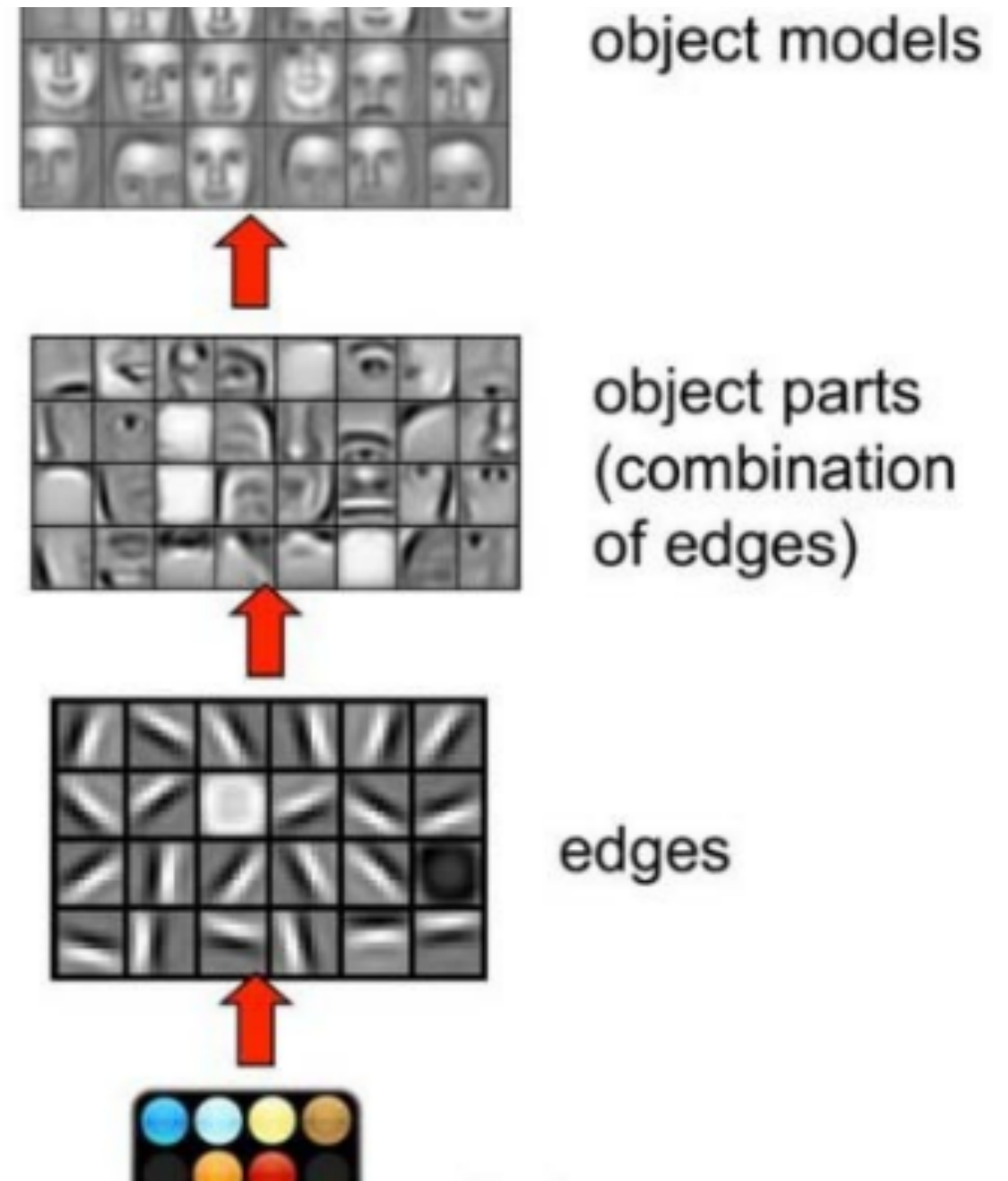
3 Submit

Our servers paint the image for you. You get an email when it's done.



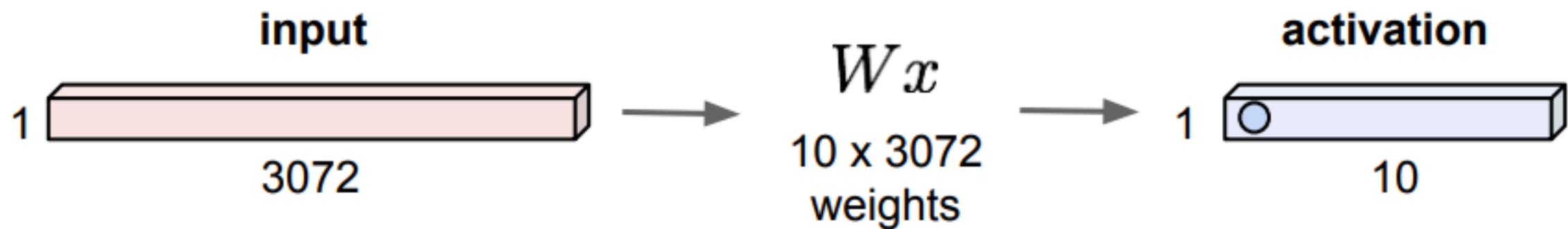
Biological Motivation

- CNNs are biologically inspired variants of DNNs.
- Our visual cortex contains a complex arrangement of cells. These cells are sensitive to sub-regions of the visual field.
- These cells act as local filters of the input space.
- Well suited to exploit strong spatially local correlation in input space.



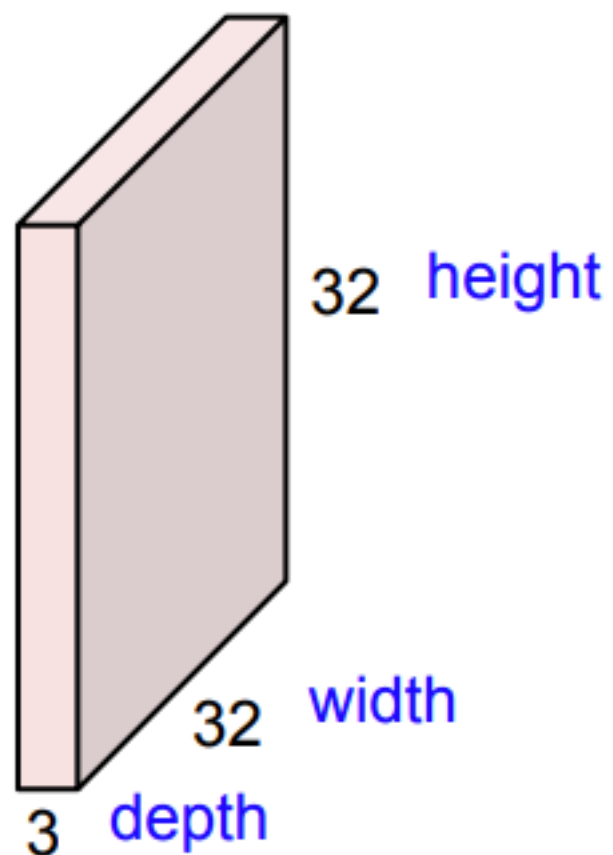
Fully Connected Layer

32x32x3 image -> stretch to 3072 x 1



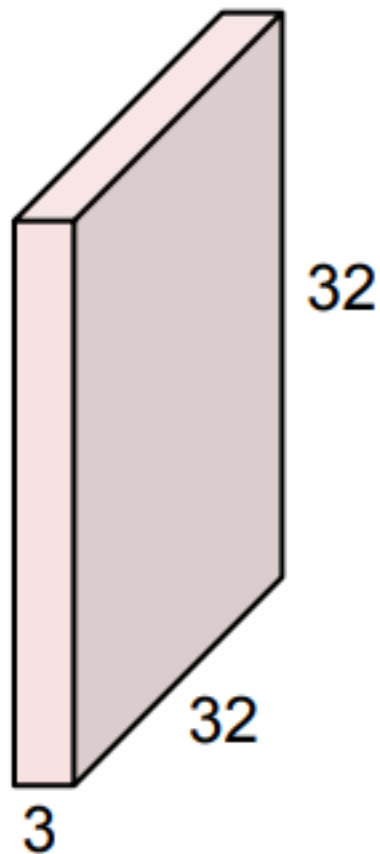
Convolutional Layer

32x32x3 image -> preserve spatial structure



Convolutional Layer

32x32x3 image

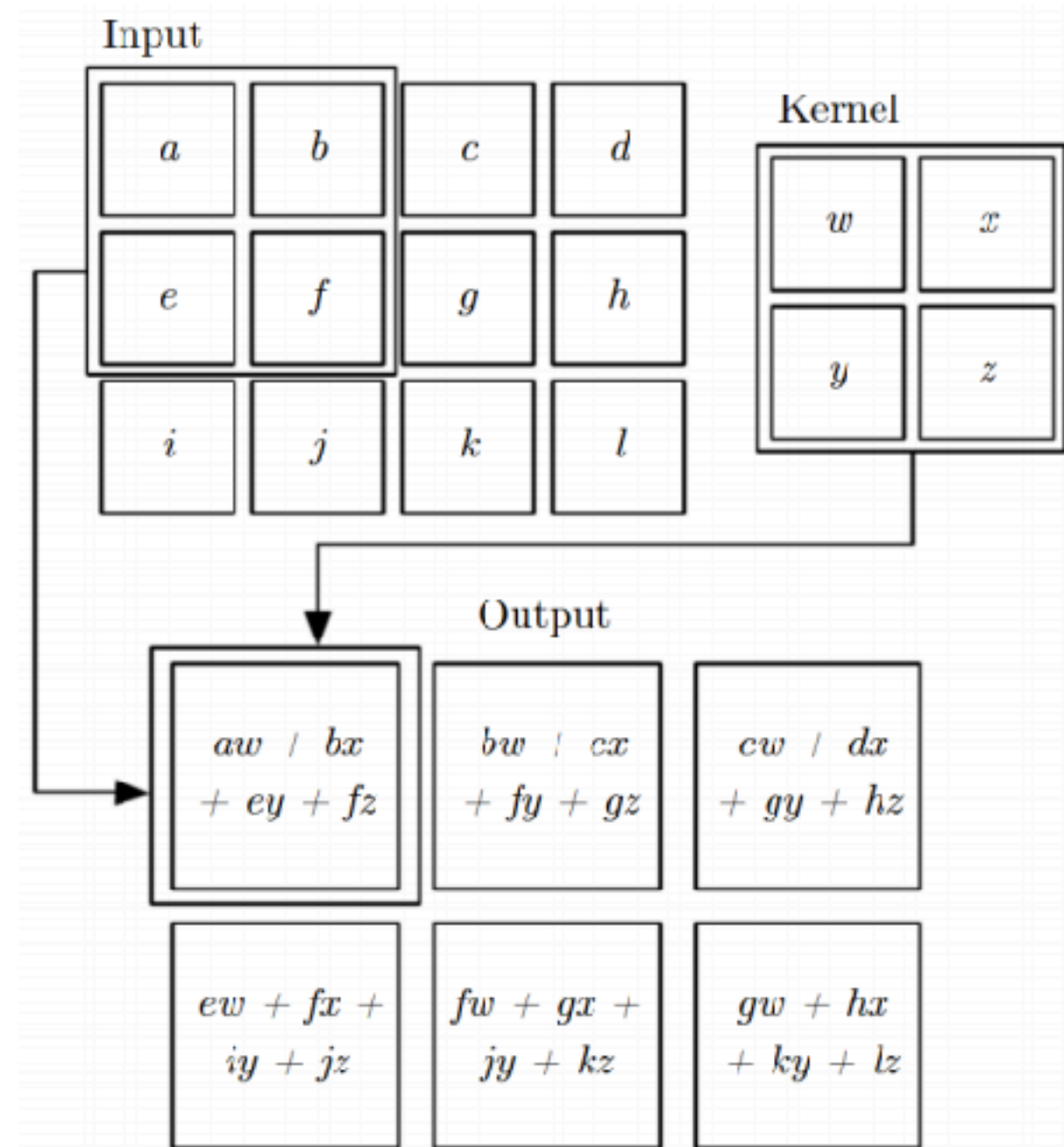


5x5x3 filter

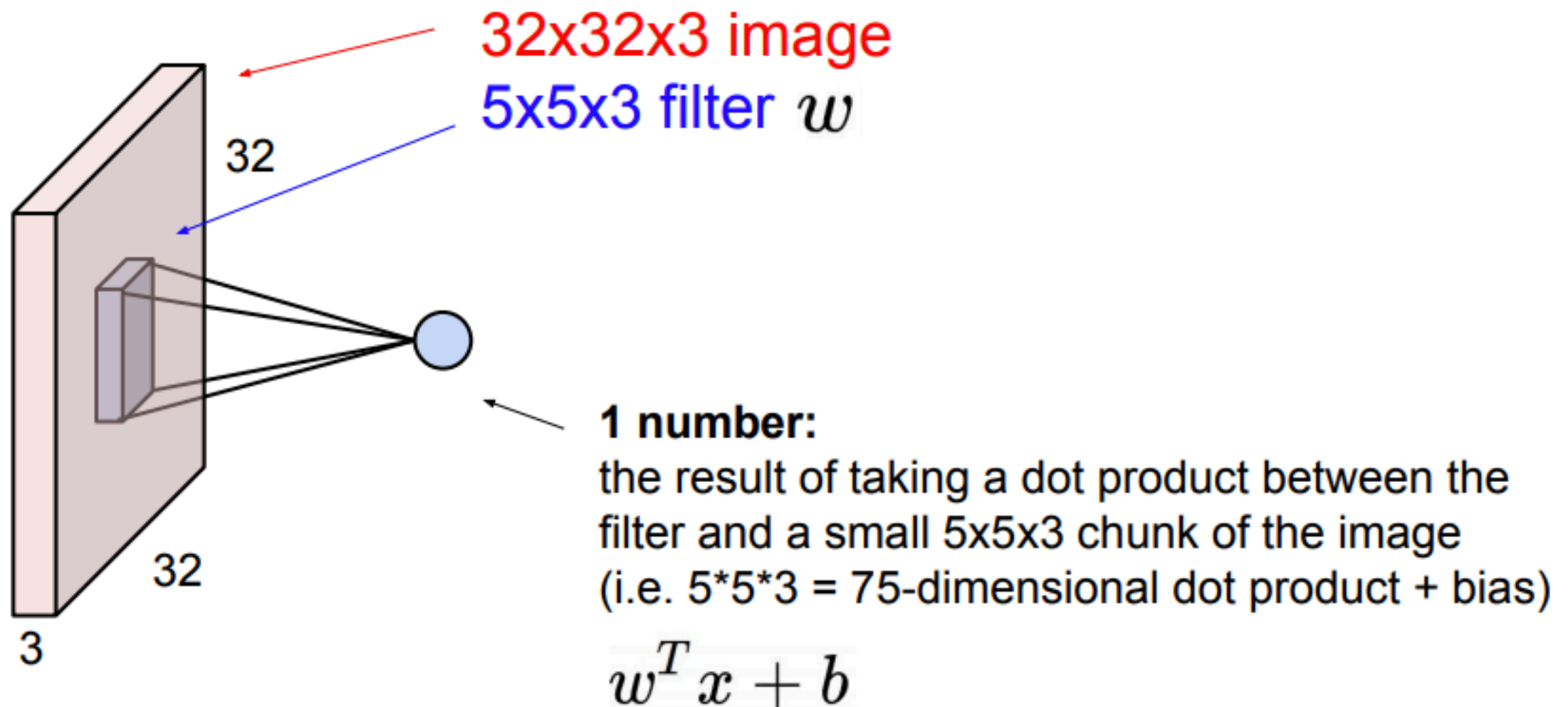


Convolve the filter with the image
i.e. “slide over the image spatially,
computing dot products”

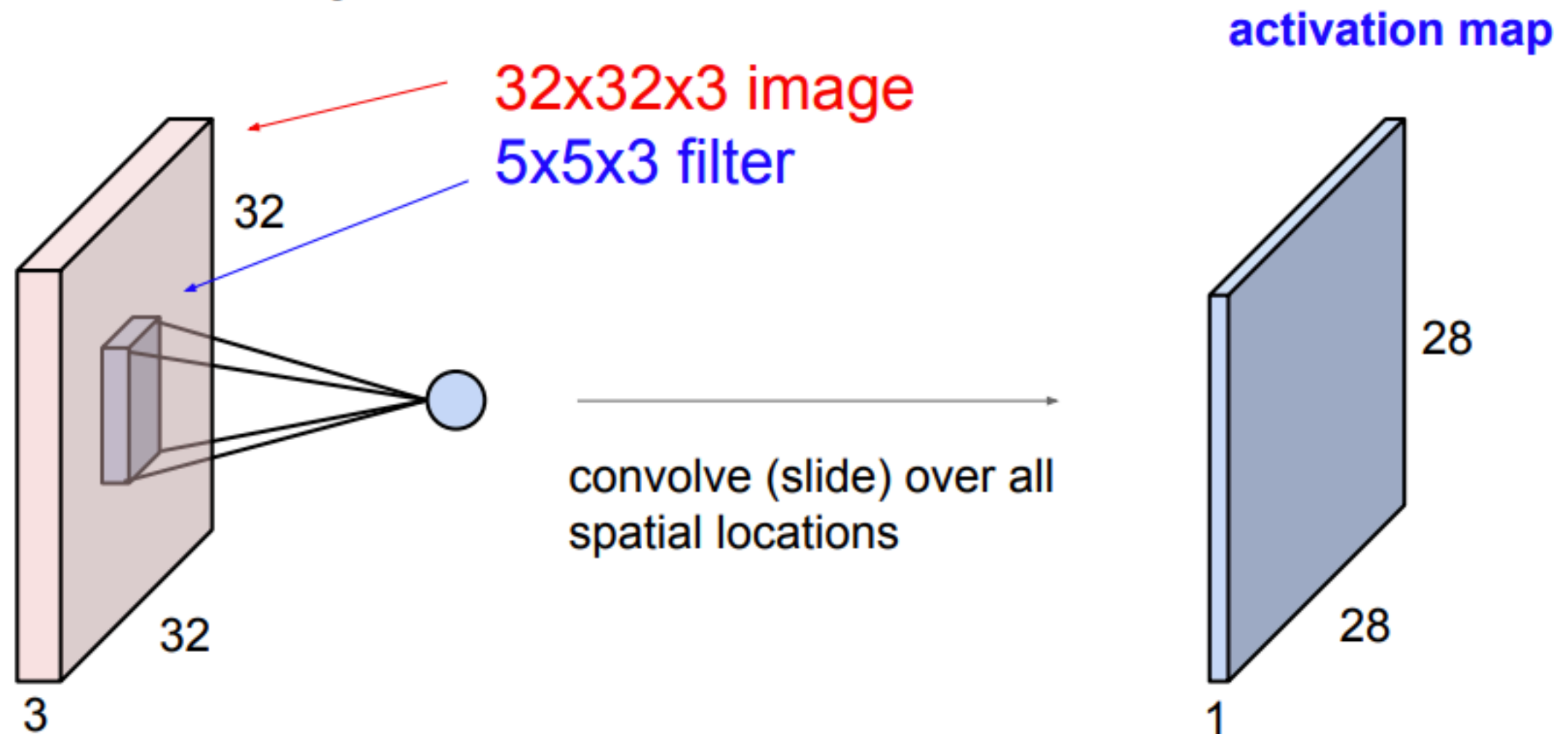
Convolutional Layer



Convolutional Layer

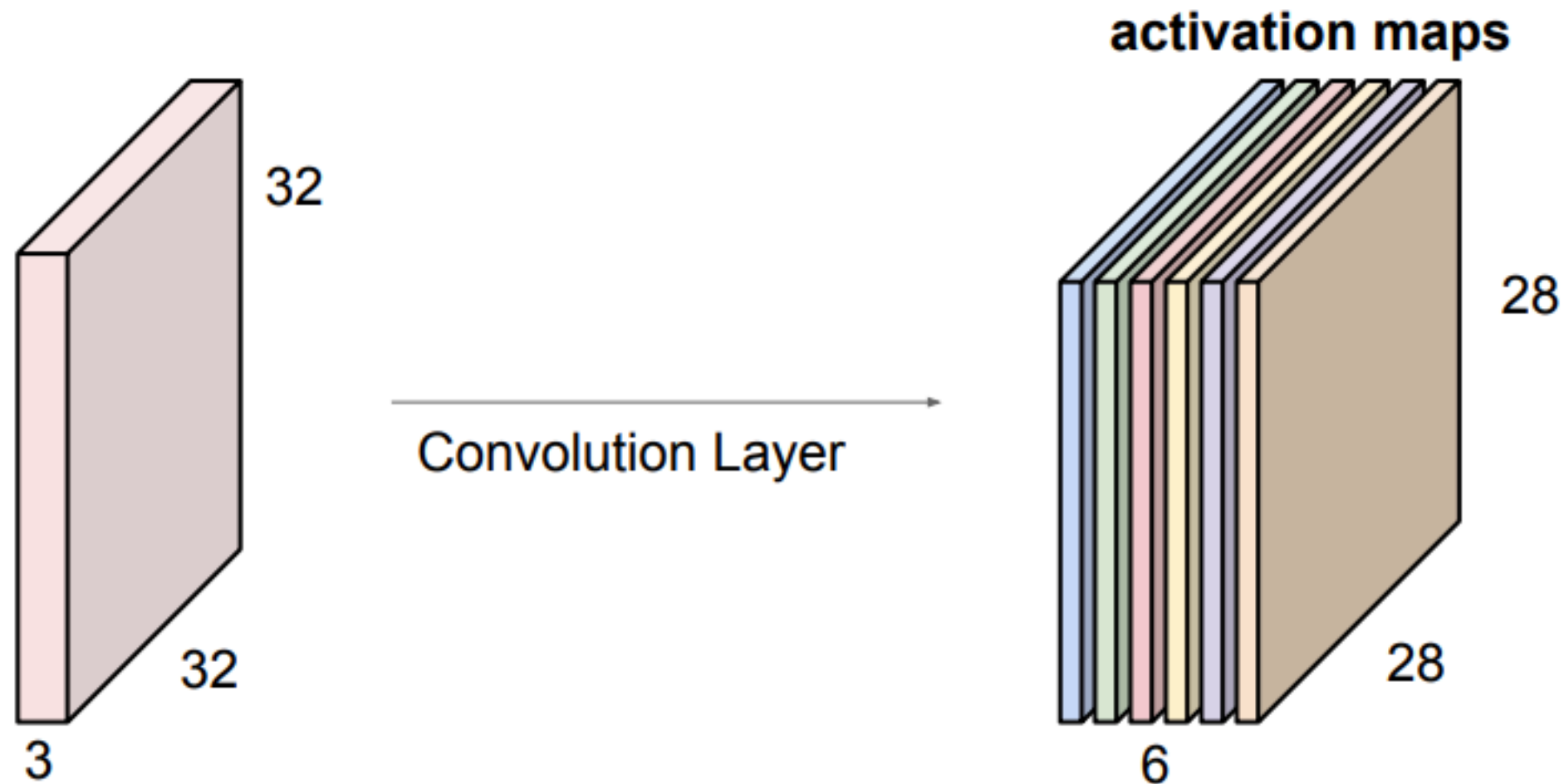


Convolutional Layer



Convolutional Layer

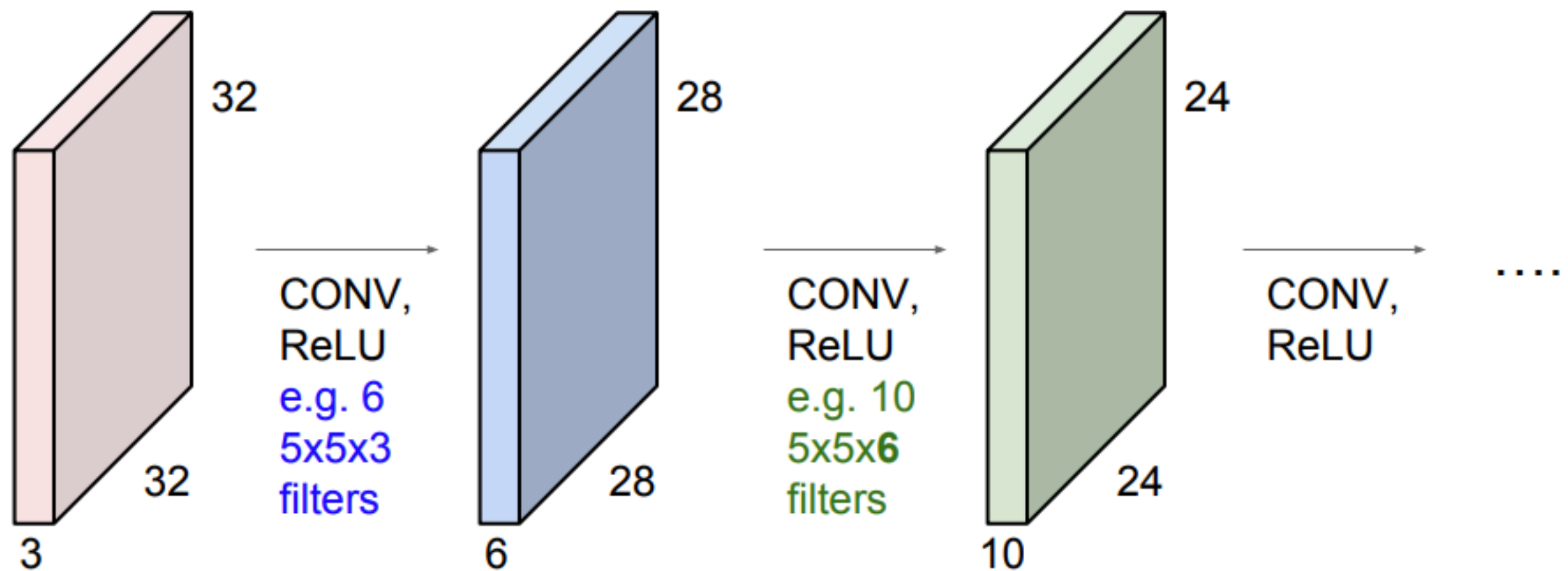
For example, if we had 6 5x5 filters, we'll get 6 separate activation maps:



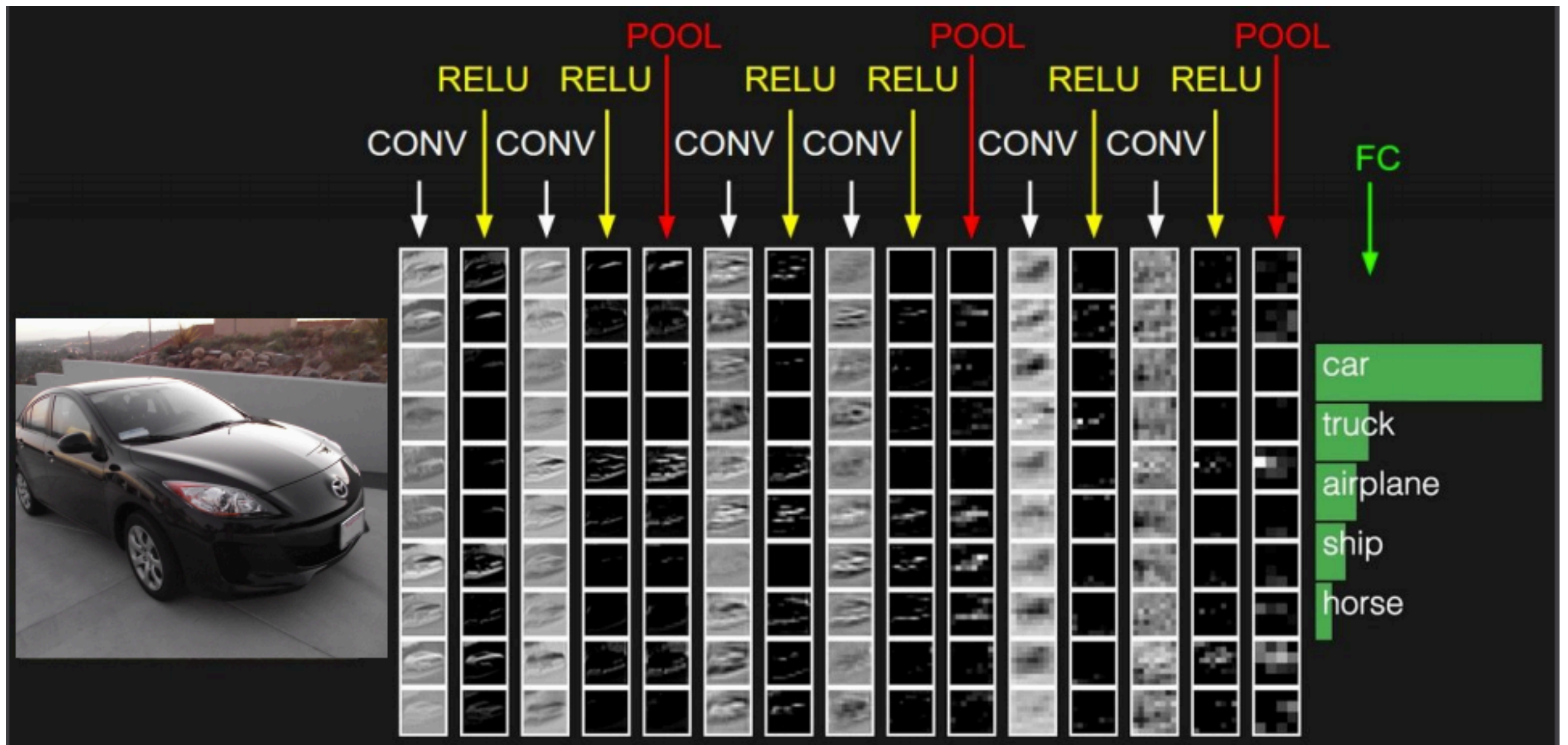
We stack these up to get a “new image” of size 28x28x6!

Convolutional Layer

Preview: ConvNet is a sequence of Convolutional Layers, interspersed with activation functions



Convolutional Layer

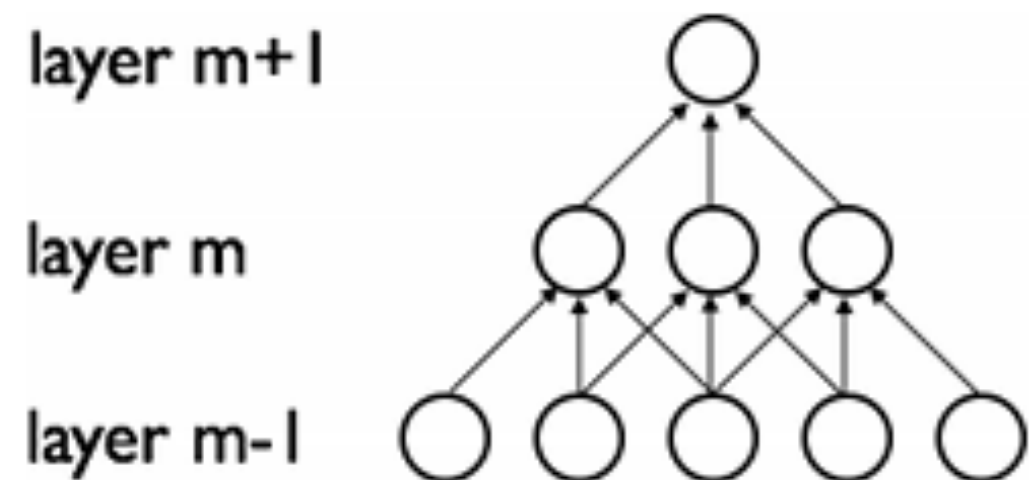


Adapted from cs231 by Fei-Fei Li & Justin Johnson & Serena Yeung

But Why?

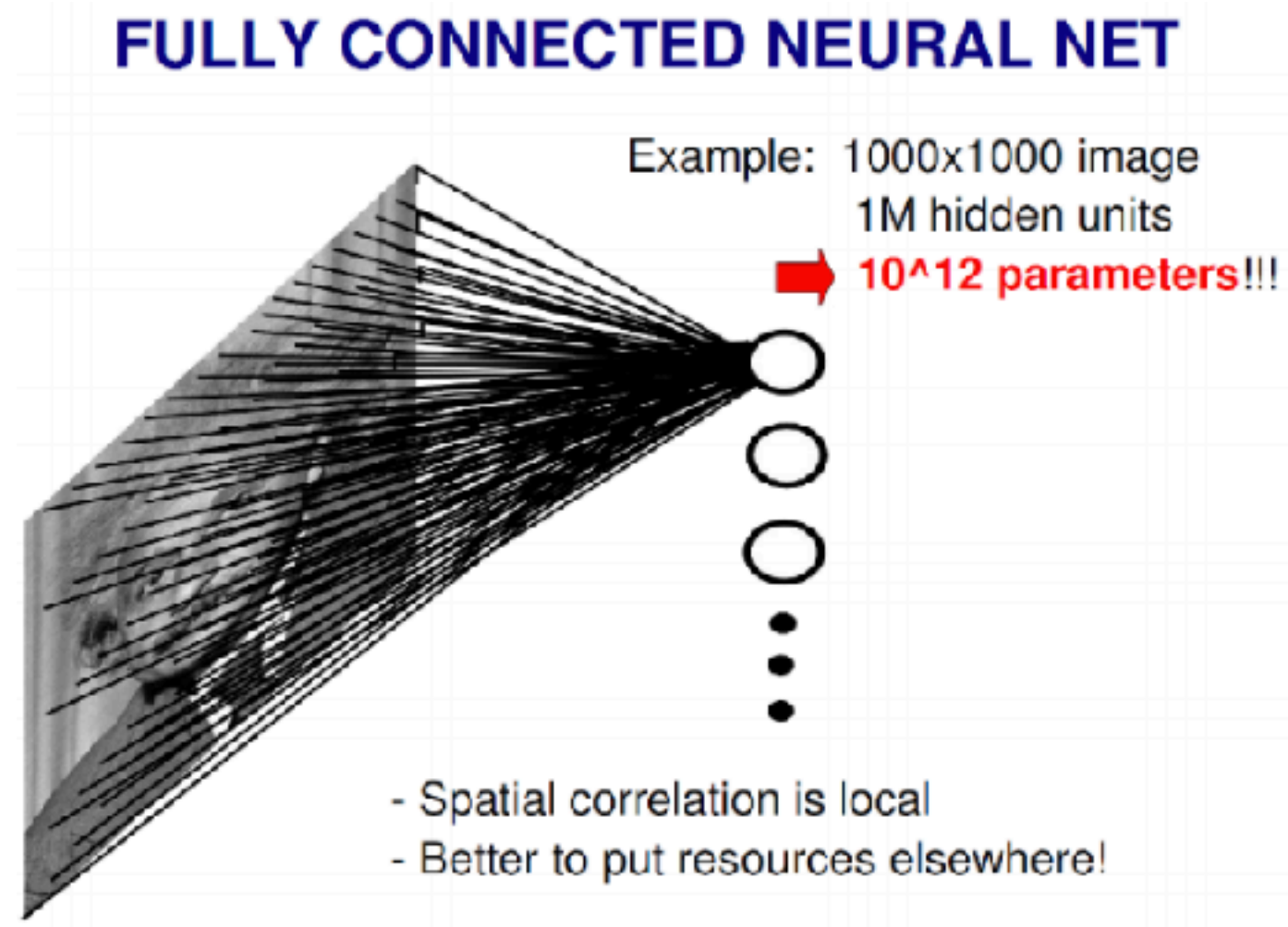
Sparse Connectivity

- The inputs of hidden units in layer **m** are from a subset of units in layer **m-1**.
- The architecture thus ensures that the learnt “filters” produce the strongest response to a spatially local input pattern.



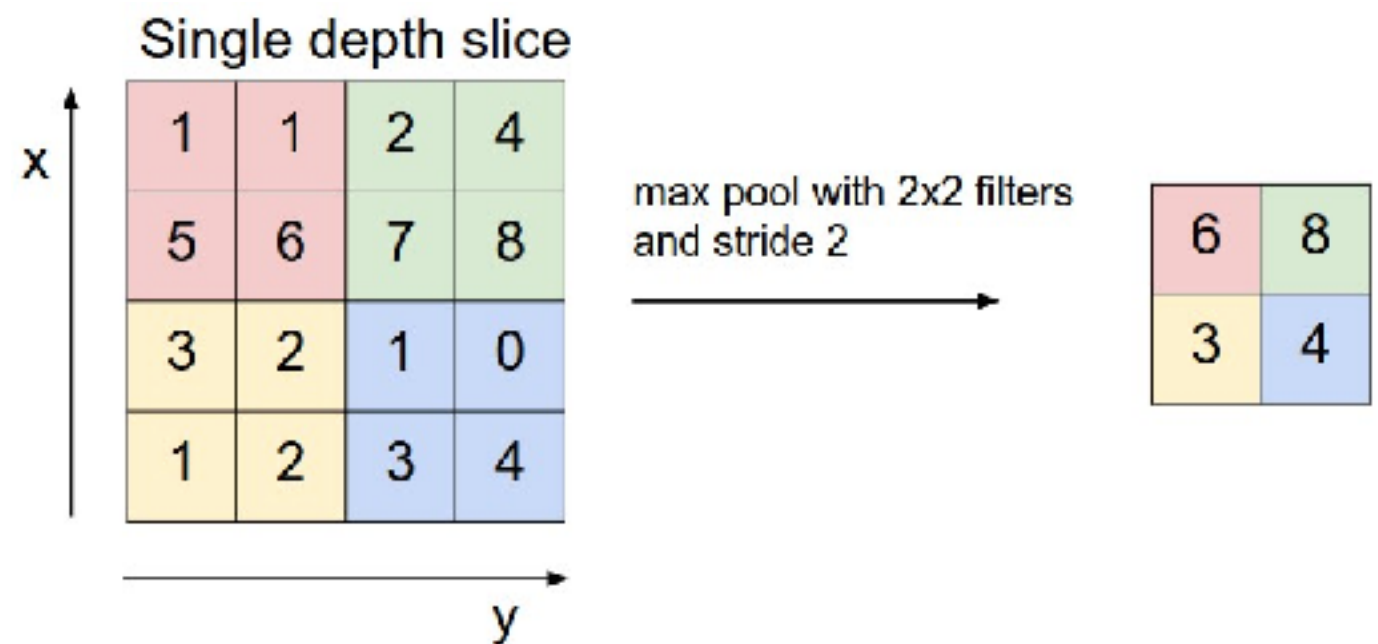
Shared Weights

- Replicating units in this way allows for features to be detected **regardless of their position in the visual field**.
- Drastic reduce in the number of free parameters compared to fully connected network **reducing overfitting** and more importantly, computational complexity of the network.

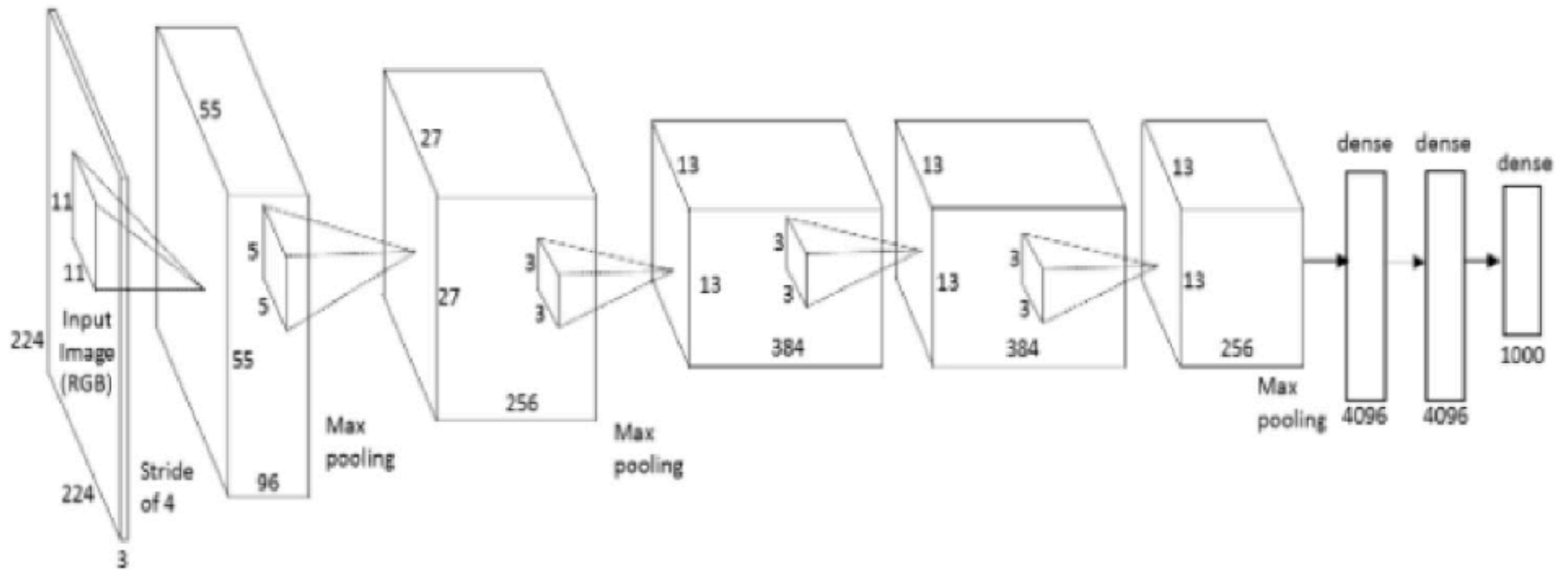


Pooling

- Non-linear down-sampling to simplify the information in output from convolutional layer.
- Variants: Max pooling (popular), Weighted average, etc.
- Useful property, if we care more about whether some feature is present than exactly where it is, thus adds robustness to position



Real CNNs



Real CNNs

