

# James Webb Space Telescope





# The landscape of “mountains” and “valleys”





# A visual grouping of five galaxies



# CONVOLUTIONAL NEURAL NETWORK

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Georgia Tech

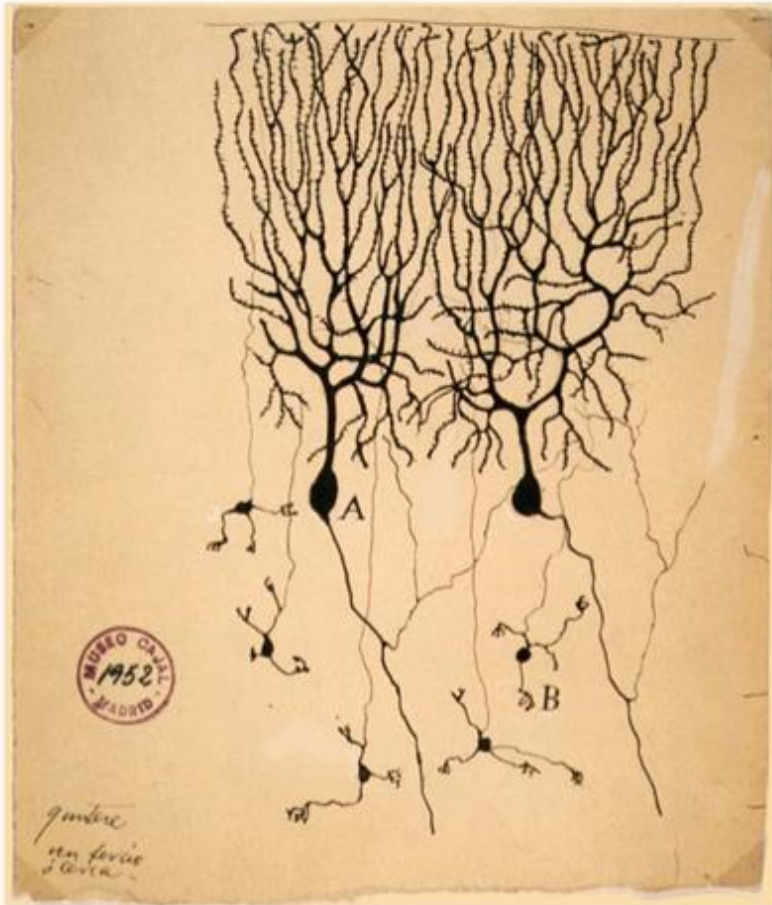
**Great visualization tool:**

**<https://poloclub.github.io/cnn-explainer/>**

Slides are based on Ming Li (University of Waterloo – Deep learning part) with some modifications



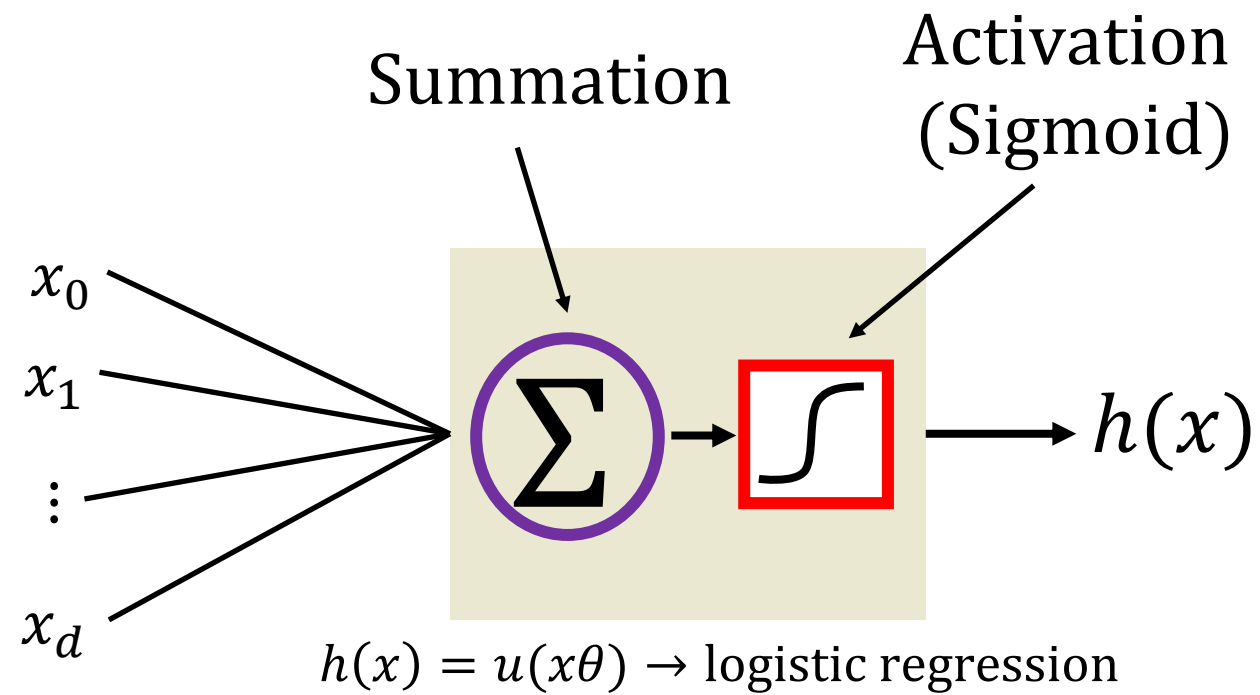
# Inspiration from Biological Neurons



The first drawing of a brain cells by Santiago Ramón y Cajal in 1899

**Neurons:** core components of brain and the nervous system consisting of

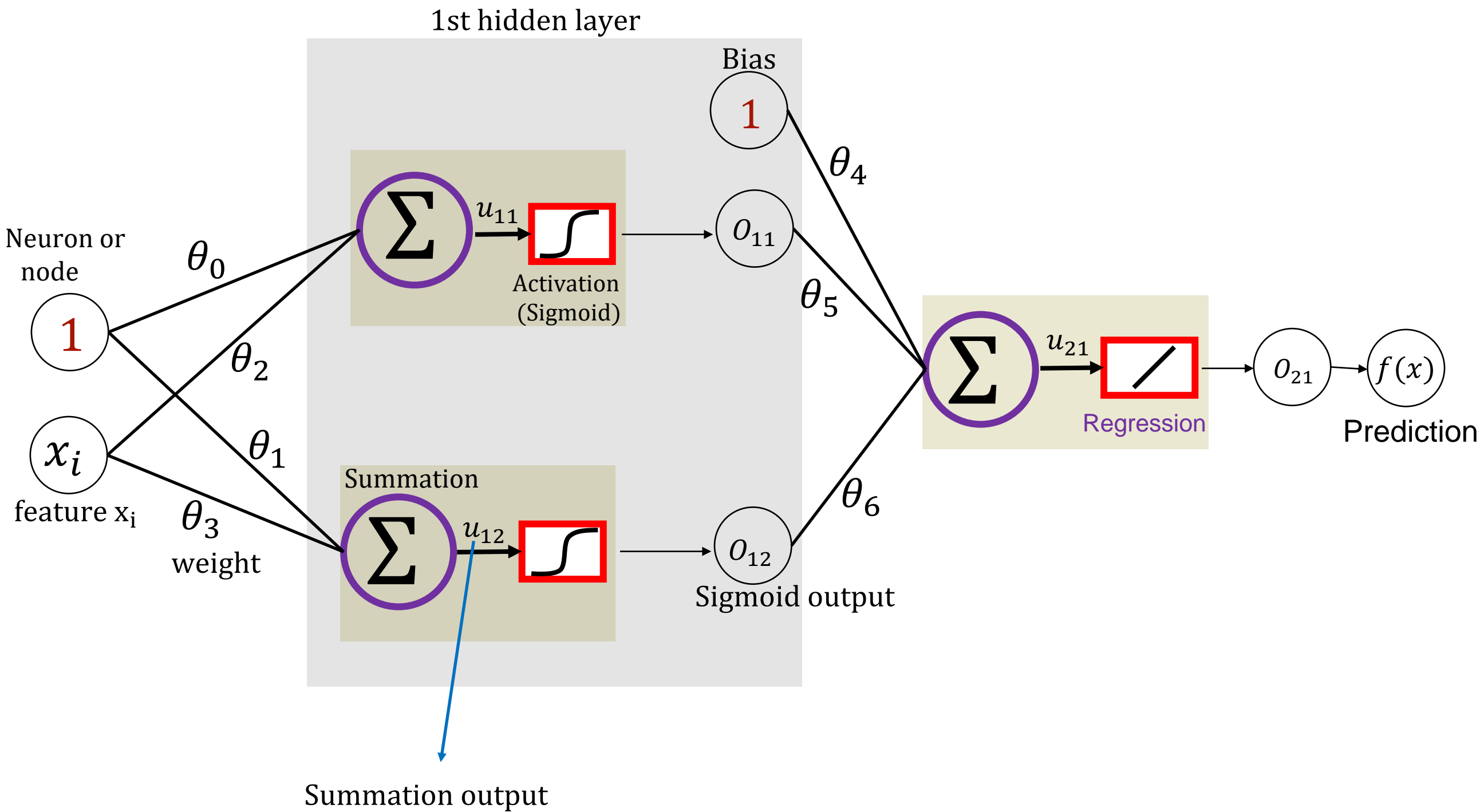
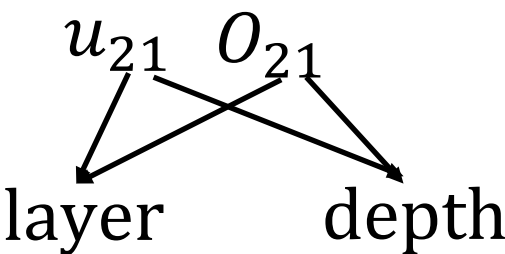
1. Dendrites that collect information from other neurons
2. An axon that generates outgoing spikes



$$\text{output} = \text{activation}(x\theta + b)$$

Name of the neuron	Activation function: $\text{activation}(z)$
Linear unit	$x\theta$
Threshold/sign unit	$\text{sign}(x\theta)$
Sigmoid unit	$\frac{1}{1 + \exp(x\theta)}$
Rectified linear unit (ReLU)	$\max(0, x\theta)$
Tanh unit	$\tanh(x\theta)$

# NN Regression

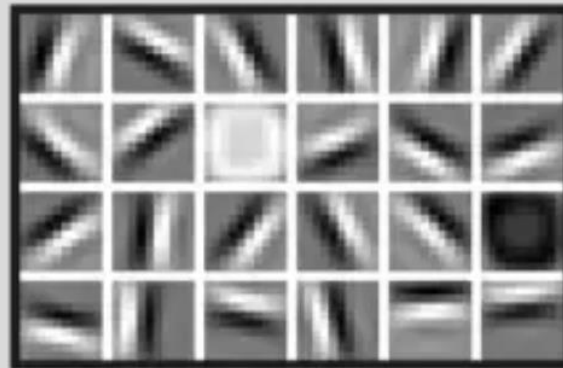


# FACIAL RECOGNITION

Deep-learning neural networks use layers of increasingly complex rules to categorize complicated shapes such as faces.



Layer 1: The computer identifies pixels of light and dark.



Layer 2: The computer learns to identify edges and simple shapes.



Layer 3: The computer learns to identify more complex shapes and objects.

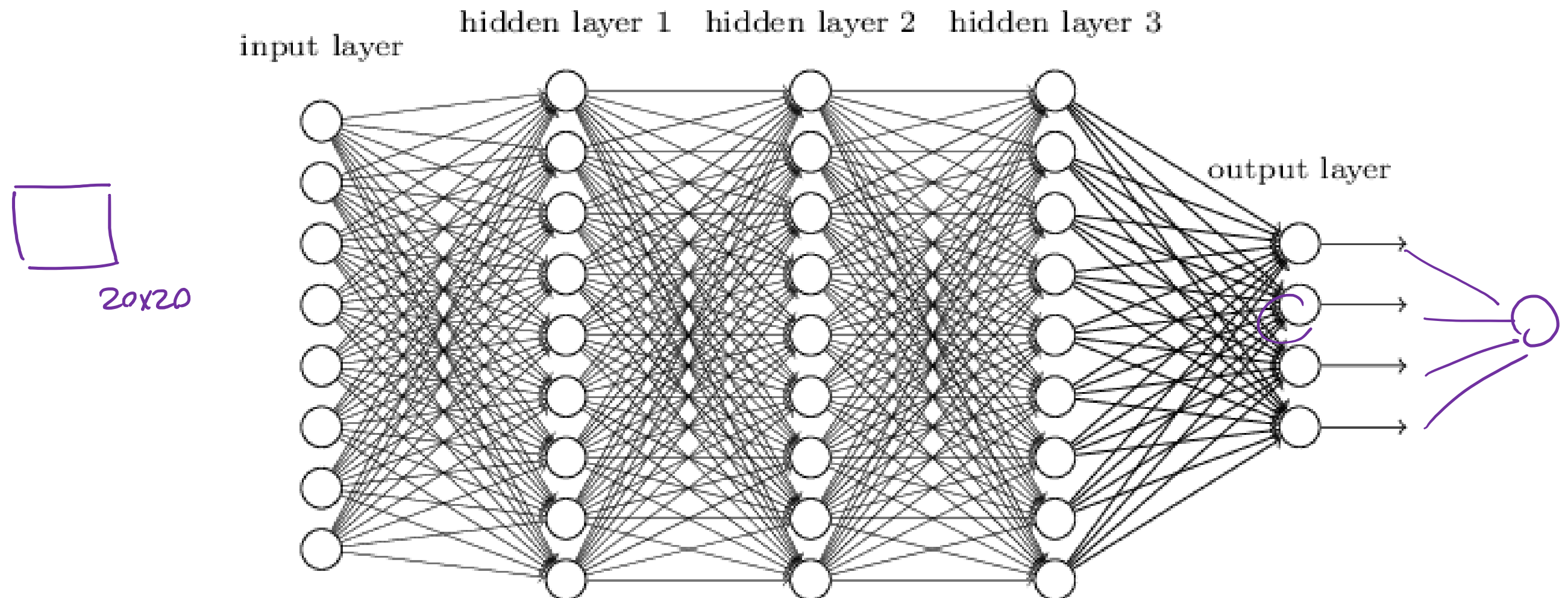


Layer 4: The computer learns which shapes and objects can be used to define a human face.



# Smaller Network: CNN

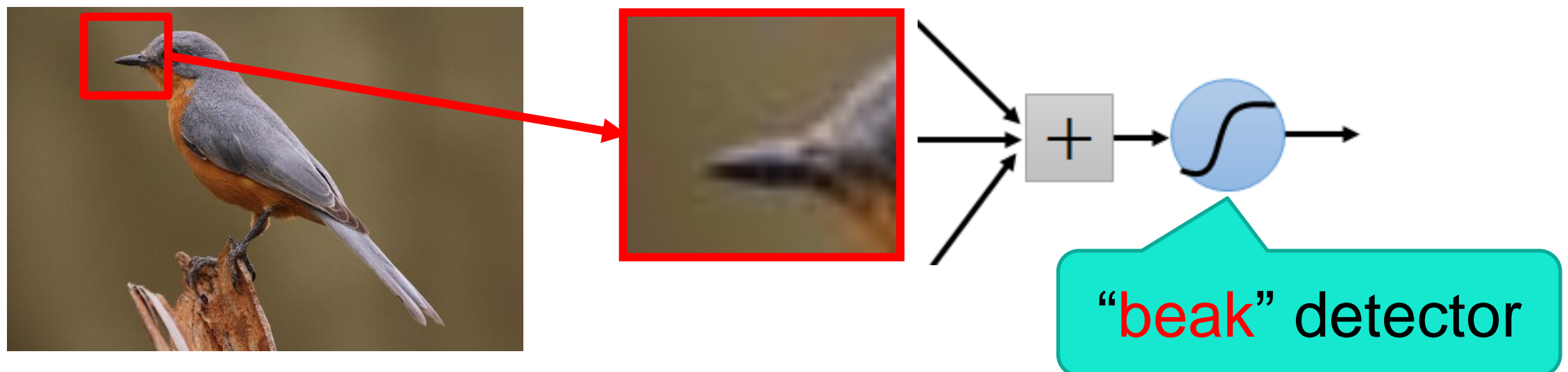
- We know it is good to learn a small model.
- From this fully connected model, do we really need all the edges?
- Can some of these be shared?



# Consider learning an image:

- Some patterns are much smaller than the whole image

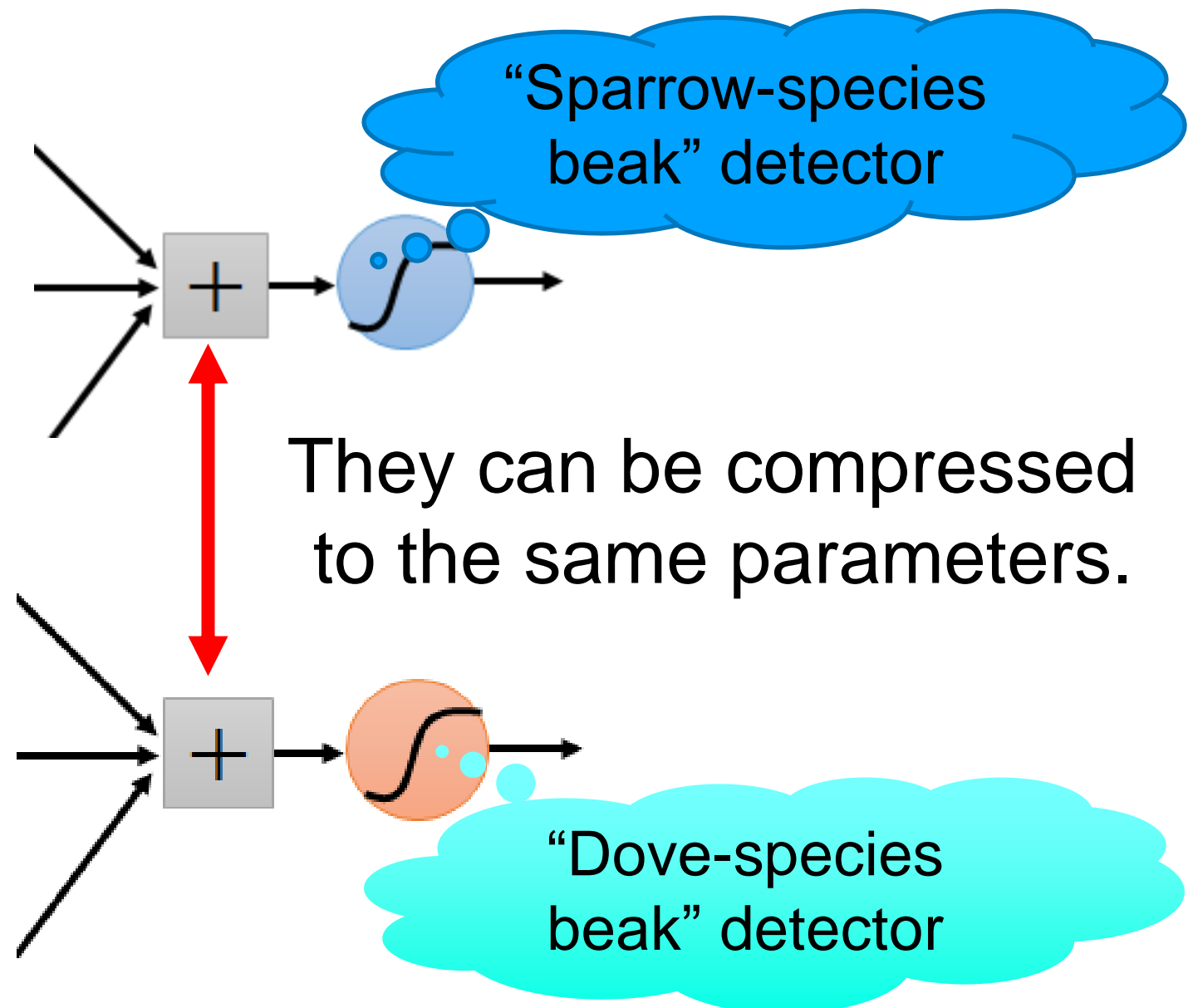
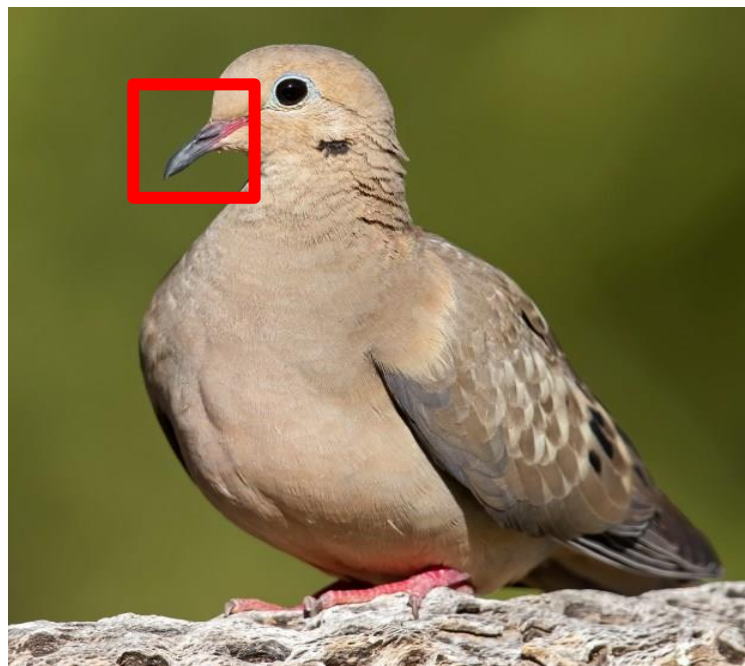
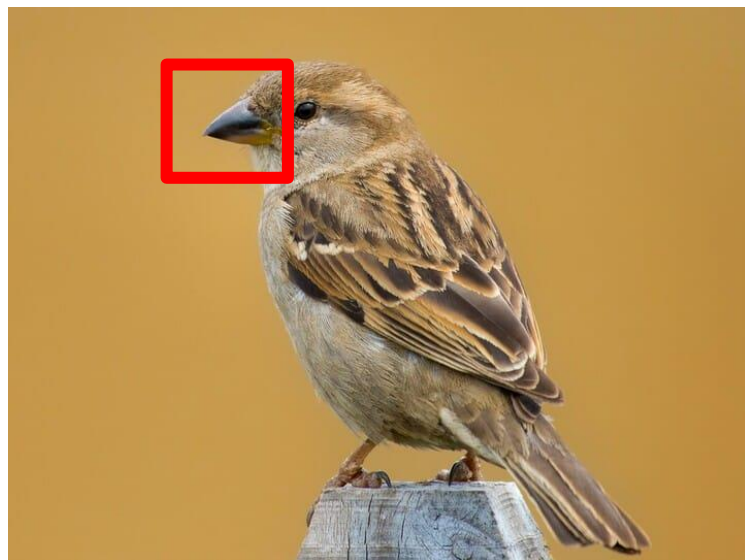
Can represent a small region with fewer parameters





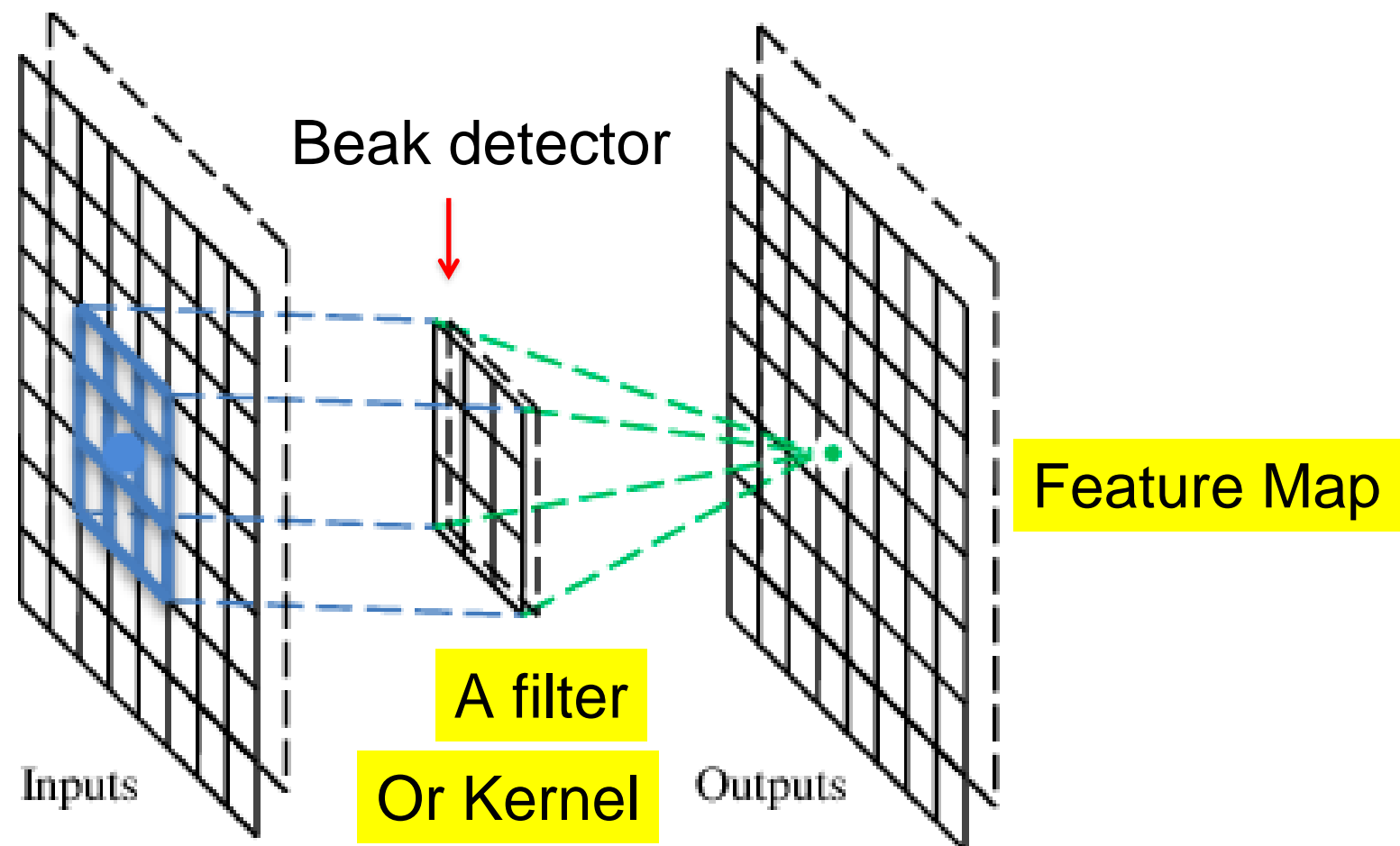
Same pattern appears in different places:  
They can be compressed!

What about training a lot of such “small” detectors  
and each detector must “move around”.



# A convolutional layer

A CNN is a neural network with some convolutional layers (and some other layers). A convolutional layer has a number of filters that does convolutional operation.





# Convolution

**These are the network parameters to be learned.**

1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

6 x 6 image

1	-1	-1
-1	1	-1
-1	-1	1

Filter 1

-1	1	-1
-1	1	-1
-1	1	-1

Filter 2

⋮ ⋮

Each filter detects a small pattern (3 x 3).

# Convolution

$$\begin{bmatrix} 1 & -1 & -1 & -1 & 1 & -1 & -1 & 1 \\ 1 & 0 & 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \rightarrow \text{sum product}$$

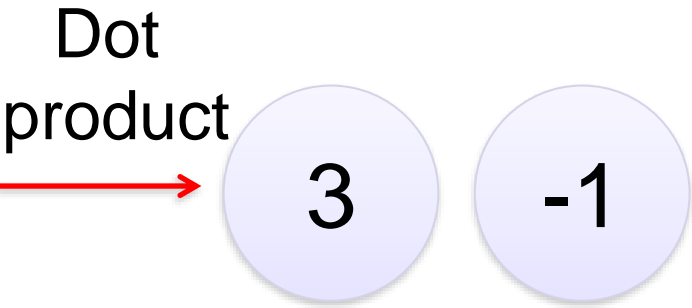
1	-1	-1
-1	1	-1
-1	-1	1

Filter 1

stride=1

1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

6 x 6 image





# Convolution

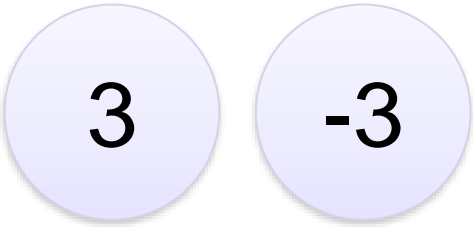
1	-1	-1
-1	1	-1
-1	-1	1

Filter 1

If stride=2

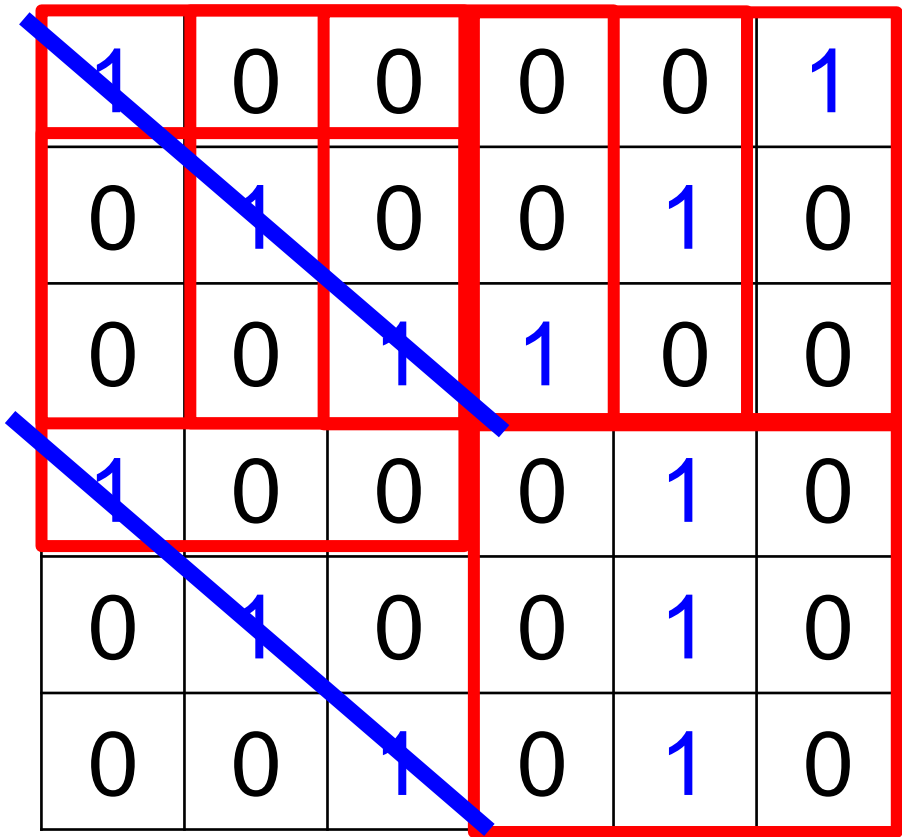
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

6 x 6 image

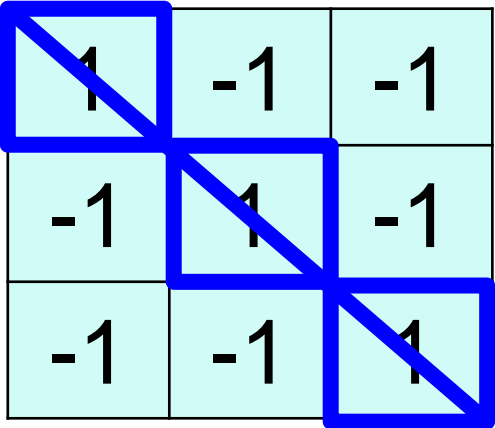


# Convolution

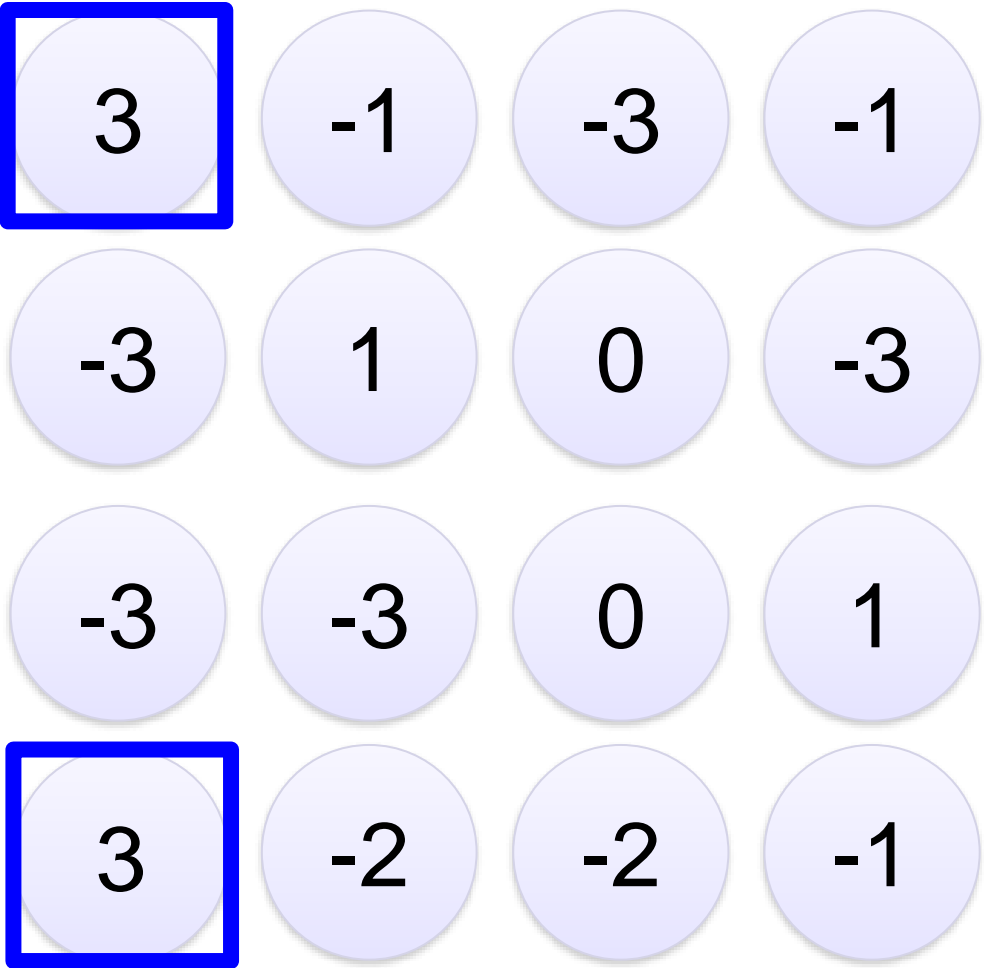
stride=1



6 x 6 image



Filter 1





# Convolution

stride=1

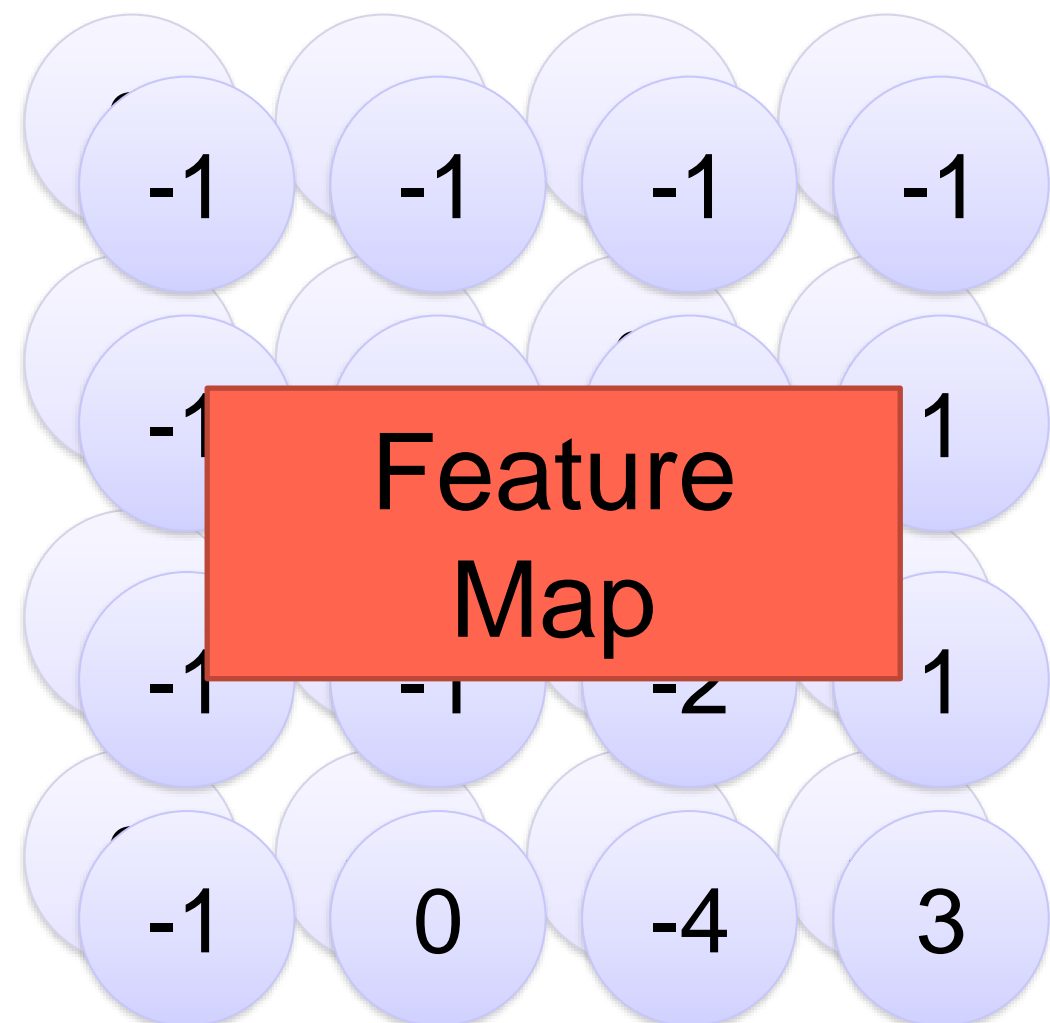
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

6 x 6 image

-1	1	-1
-1	1	-1
-1	1	-1

Filter 2

Repeat this for each filter



Two 4 x 4 images  
Forming 2 x 4 x 4 matrix

# Color image: RGB 3 channels

Sub-matrix  $\begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 1 & \dots \end{bmatrix}$  length = 27

Filter  $\begin{bmatrix} 1 & -1 & -1 & -1 & 1 & -1 & -1 & -1 & 1 & 1 & \dots \end{bmatrix}$  length = 27

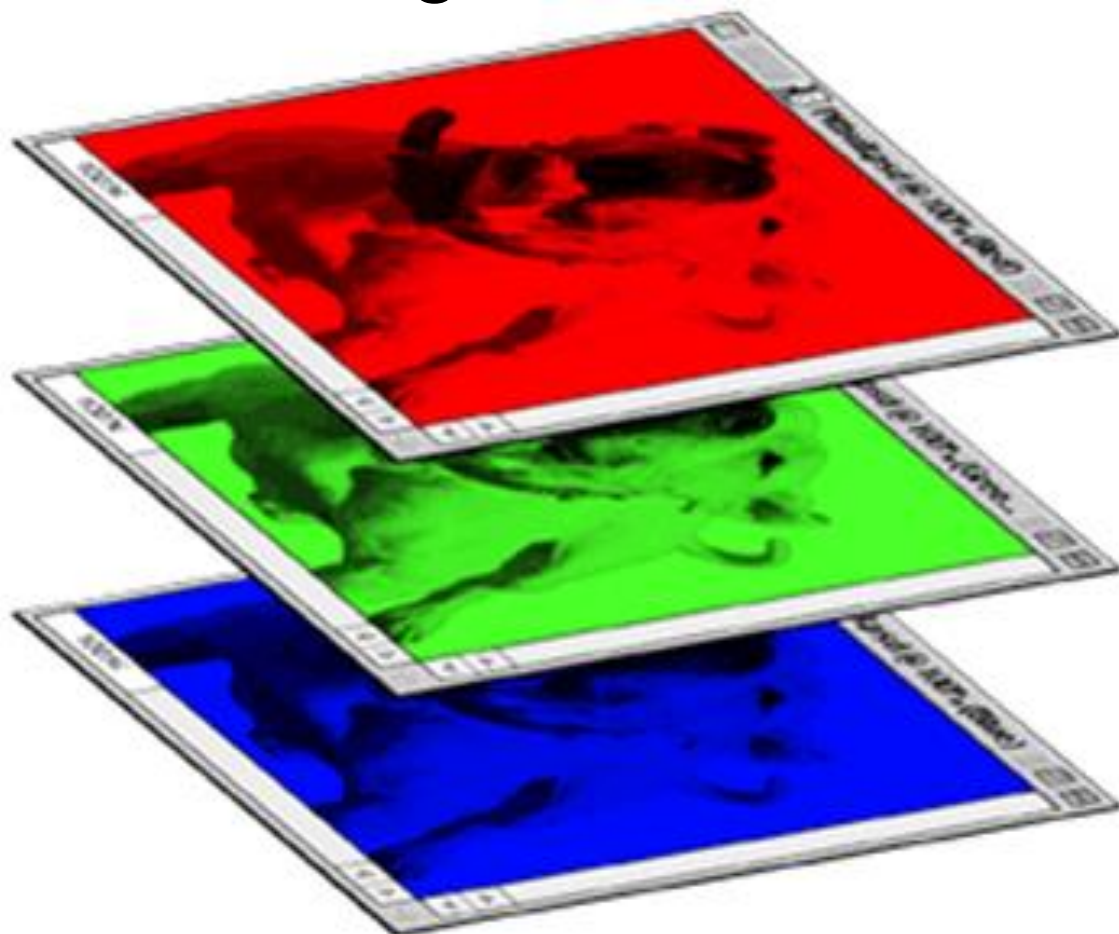
1	-1	-1
-1	1	-1
-1	-1	1

Filter 1

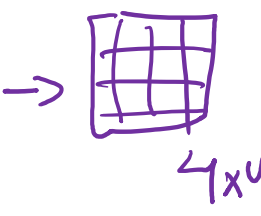
-1	1	-1
-1	1	-1
-1	1	-1

Filter 2

Color image



1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0





# Convolution v.s. Fully Connected

1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

image

1	-1	-1
-1	1	-1
-1	-1	1

-1	1	-1
-1	1	-1
-1	1	-1

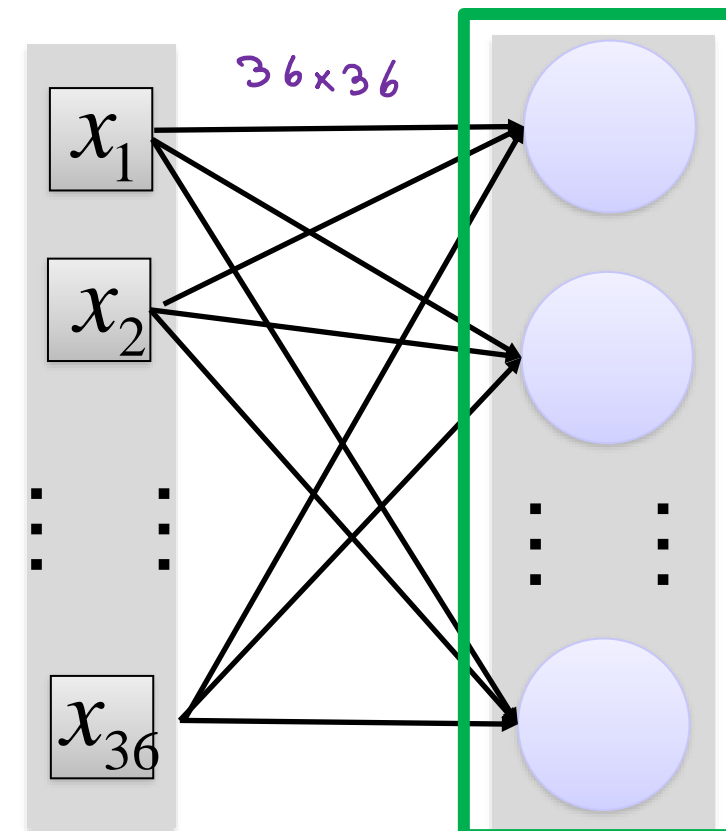


convolution

-1	-1	-1	-1
-1	-1	-2	1
-1	-1	-2	1
-1	0	-4	3

Fully-  
connected

1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0



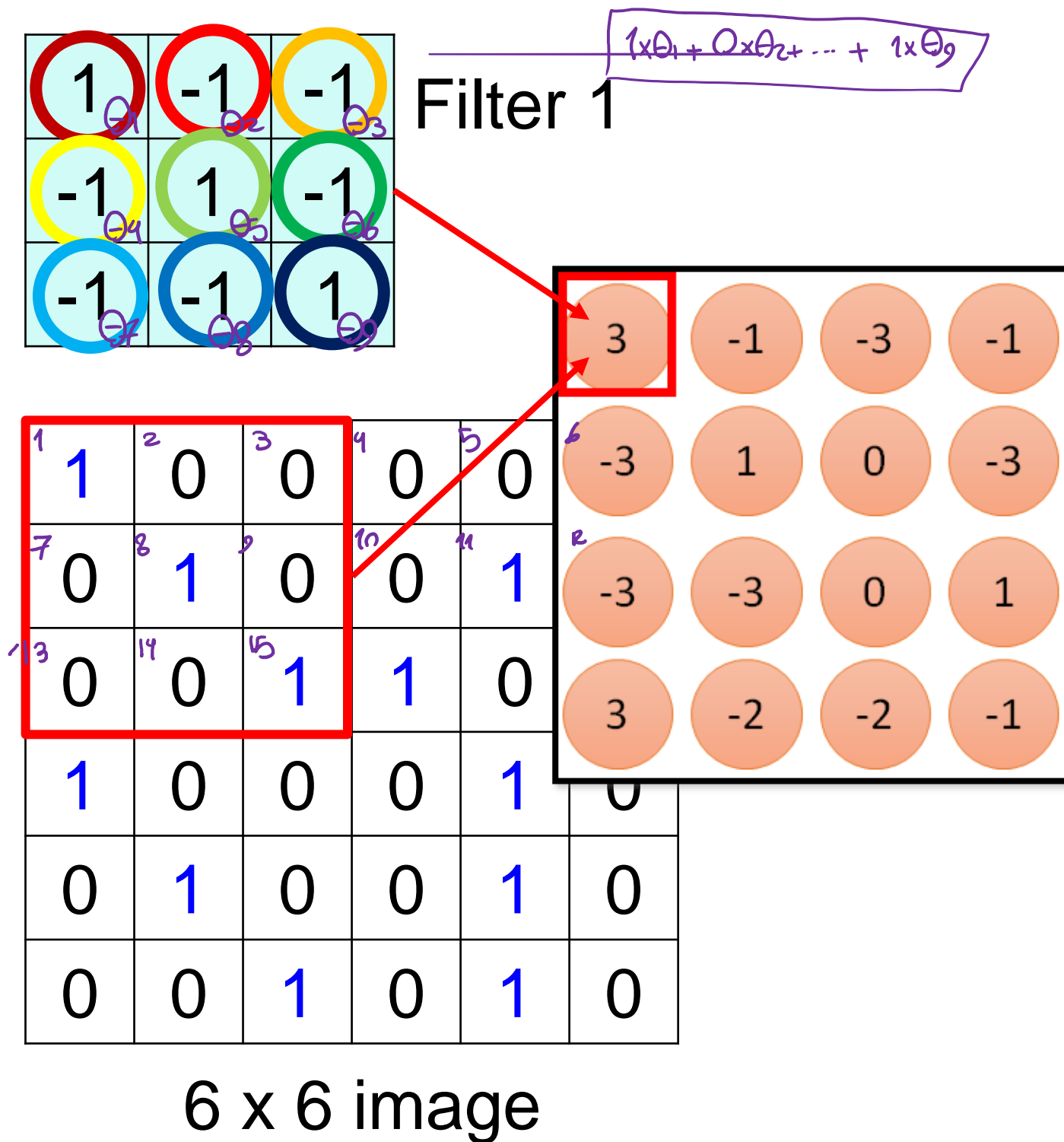
Conventional  
Fully Connected  
layers  
(FC layers)

1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

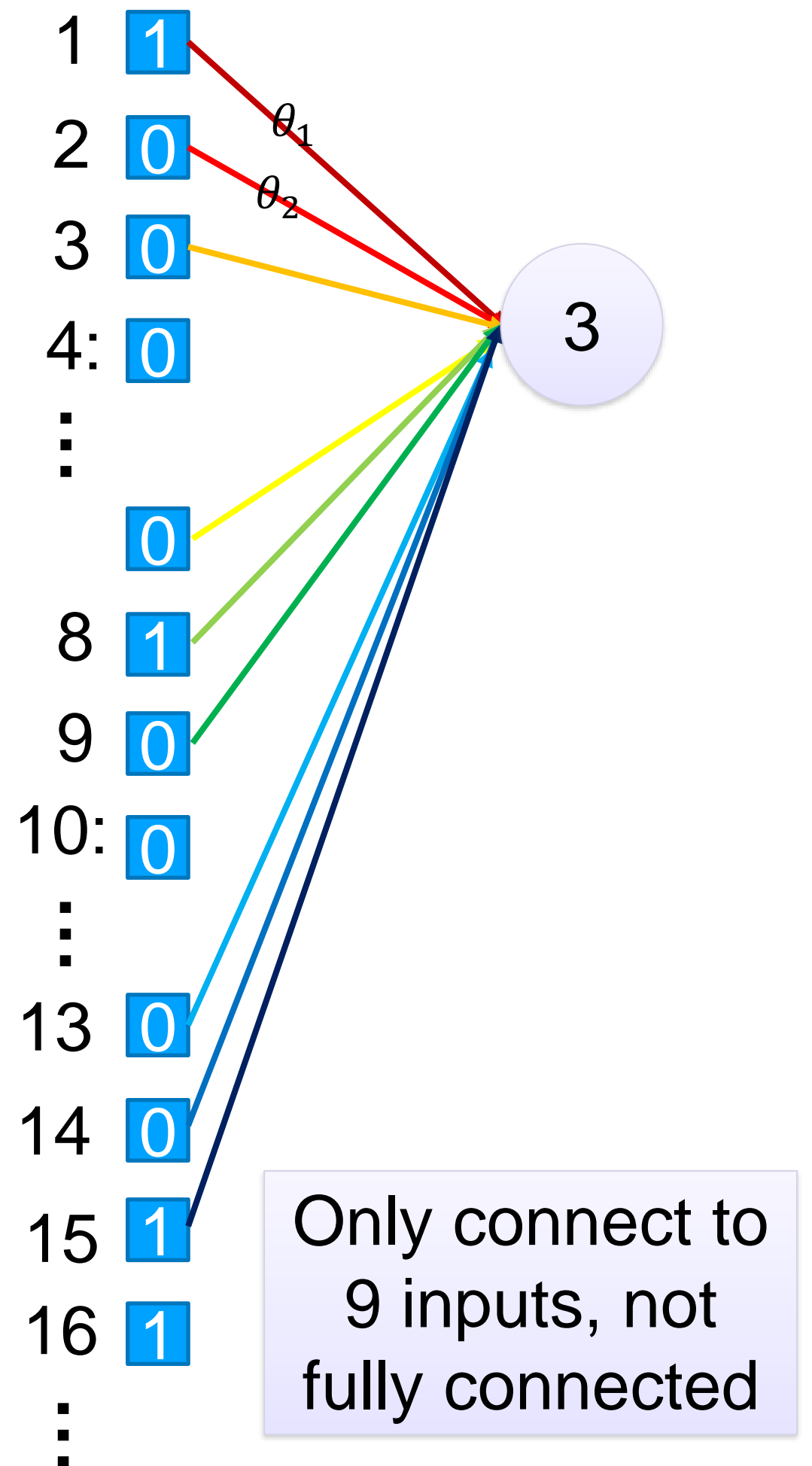
6 x 6 image

1	1	
2	0	
3	0	
4	0	
5	0	
6	1	
7	0	
8	1	
⋮		⋮
31	0	
32	0	
33	1	
34	0	
35	1	
36	0	

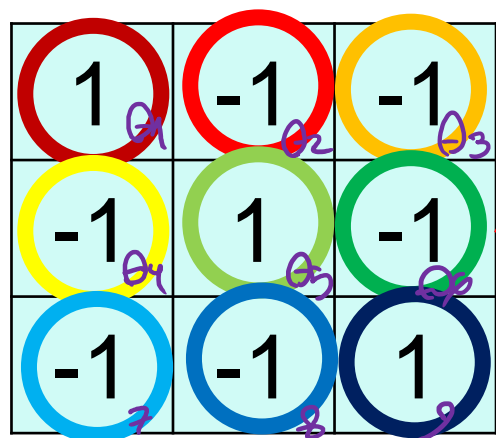
features    1<sup>st</sup> hidden layer



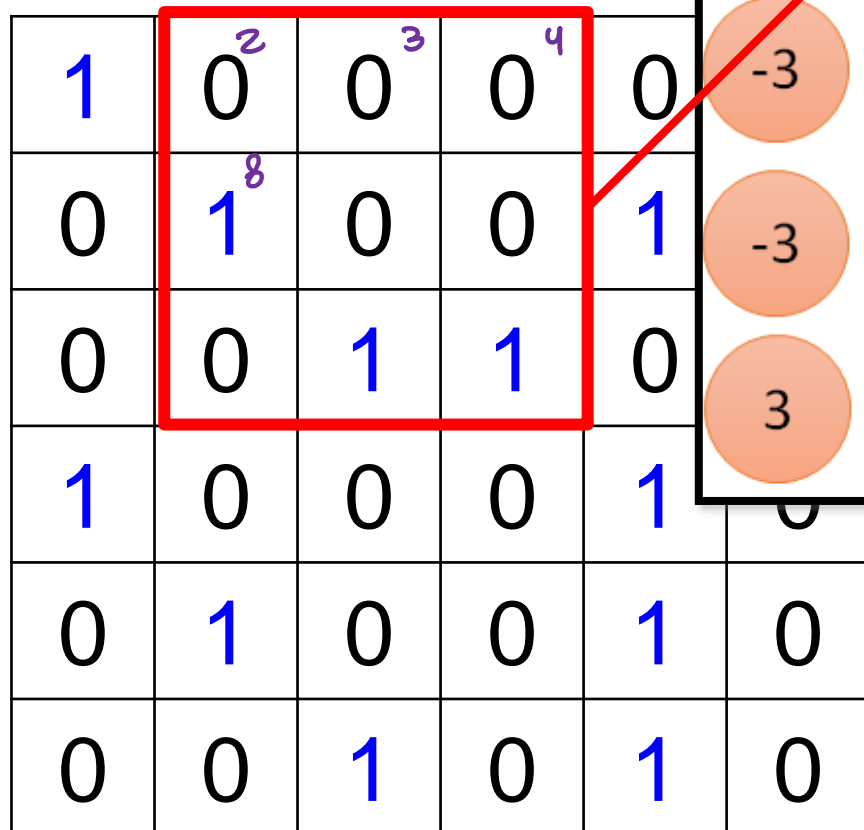
fewer parameters!







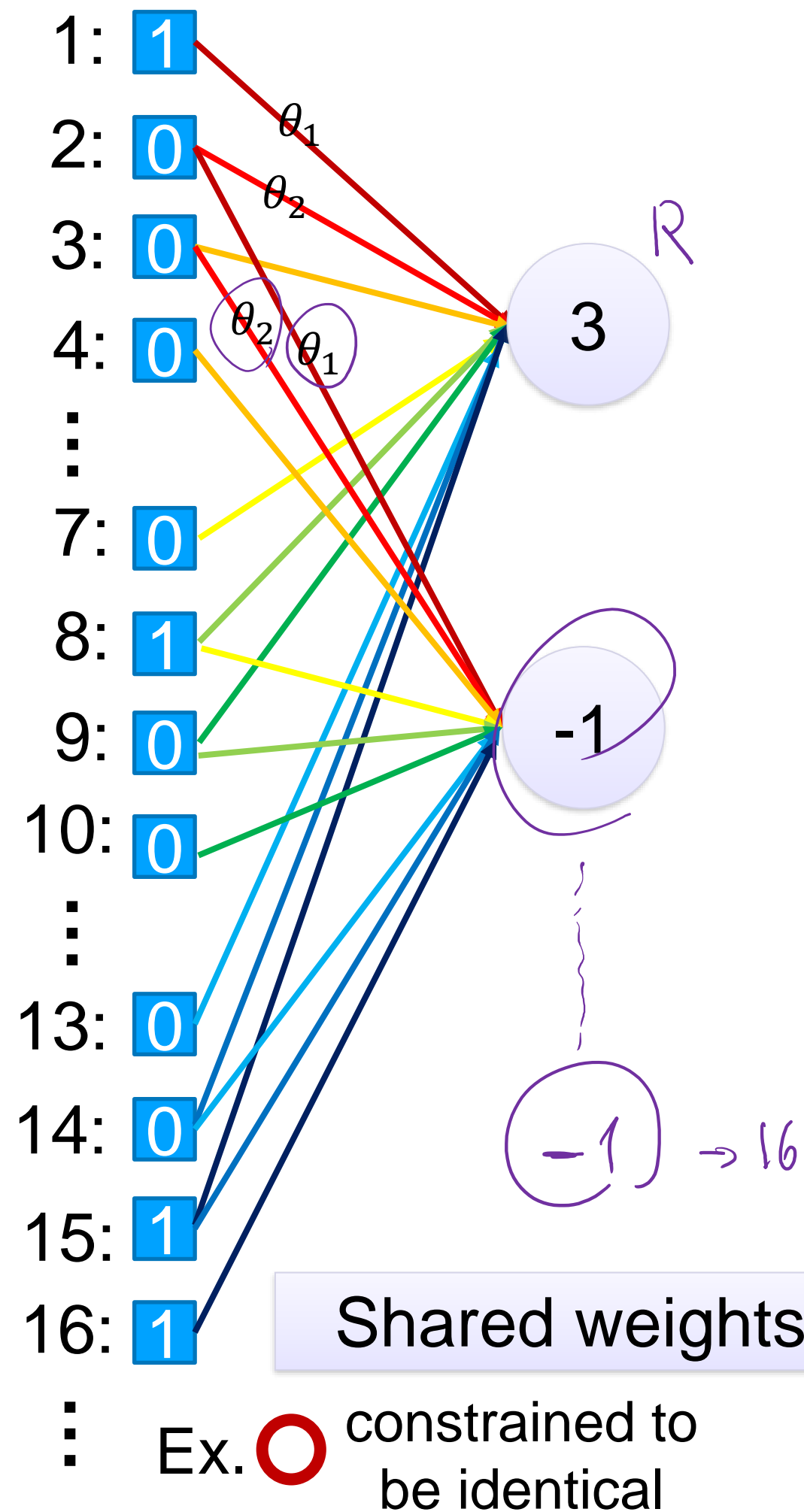
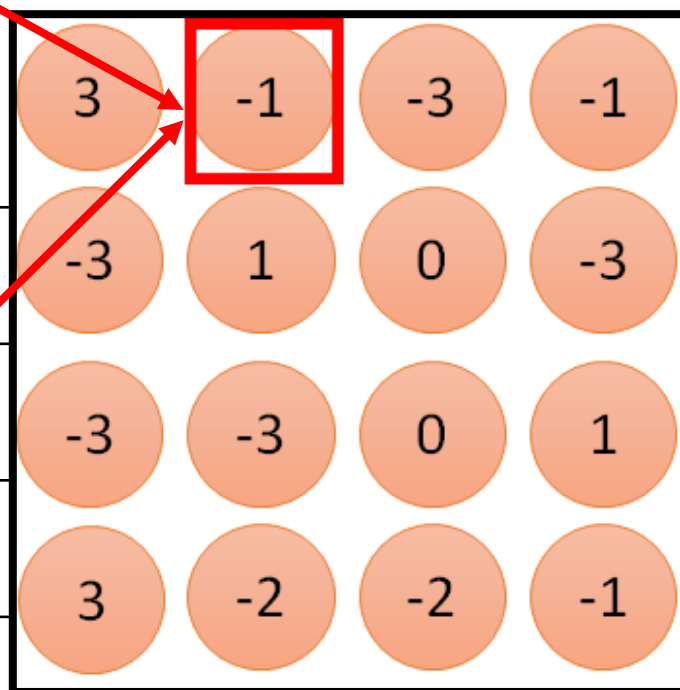
Filter 1



6 x 6 image

Fewer parameters

Even fewer parameters

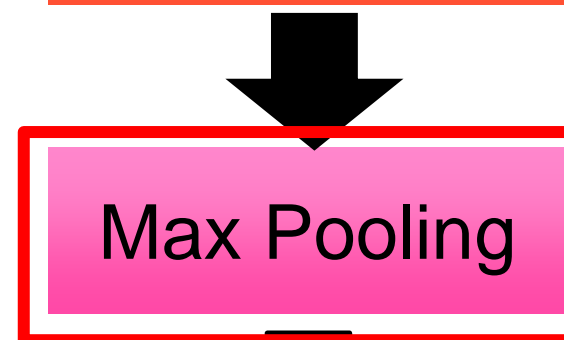
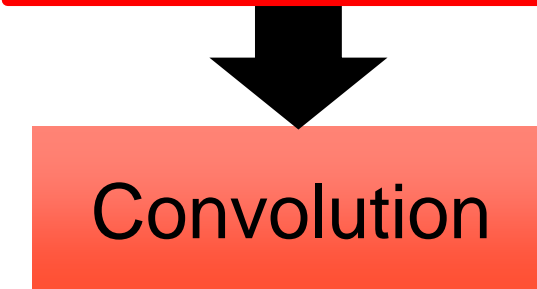
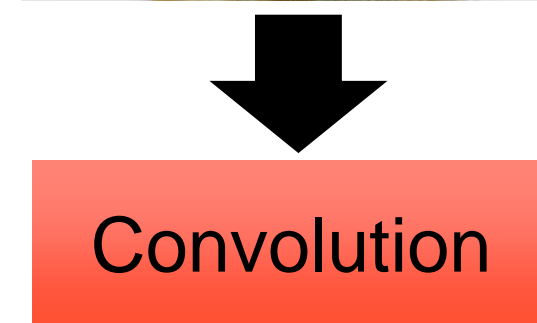
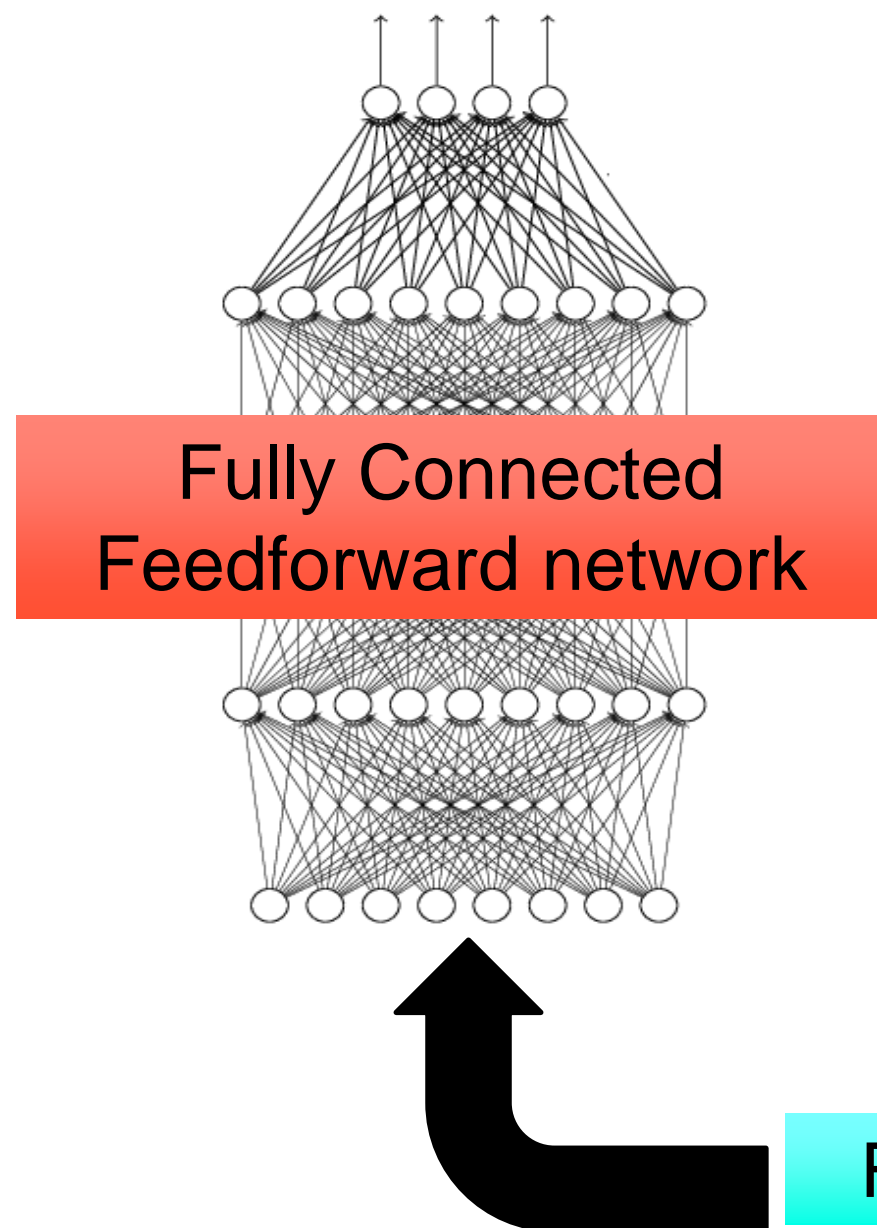


Shared weights

Ex. ○ constrained to be identical

# The whole CNN

cat dog .....



Can repeat many times

Flattened



# Max Pooling

1	-1	-1
-1	1	-1
-1	-1	1

Filter 1

$$\begin{bmatrix} 3 & 0 \\ 3 & 1 \end{bmatrix}_{2 \times 2}$$

3	-1	-3	-1
-3	1	0	-3
-3	-3	0	1
3	-2	-2	-1

-1	1	-1
-1	1	-1
-1	1	-1

Filter 2

$$\begin{bmatrix} -1 & 1 \\ 0 & 3 \end{bmatrix}_{2 \times 2}$$

-1	-1	-1	-1
-1	-1	-2	1
-1	-1	-2	1
-1	0	-4	3



# Why Pooling

- Subsampling pixels will not change the object  
bird

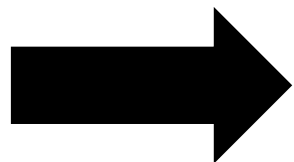


Subsampling



bird

We can subsample the pixels to make image smaller



fewer parameters to characterize the image

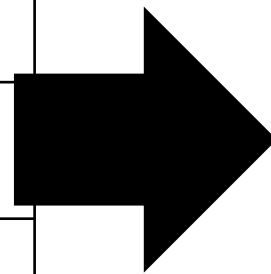
# A CNN compresses a fully connected network in three ways:

- Reducing number of connections
- Shared weights on the edges
- Max pooling further reduces the complexity

# Max Pooling

1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

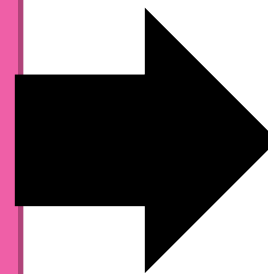
6 x 6 image



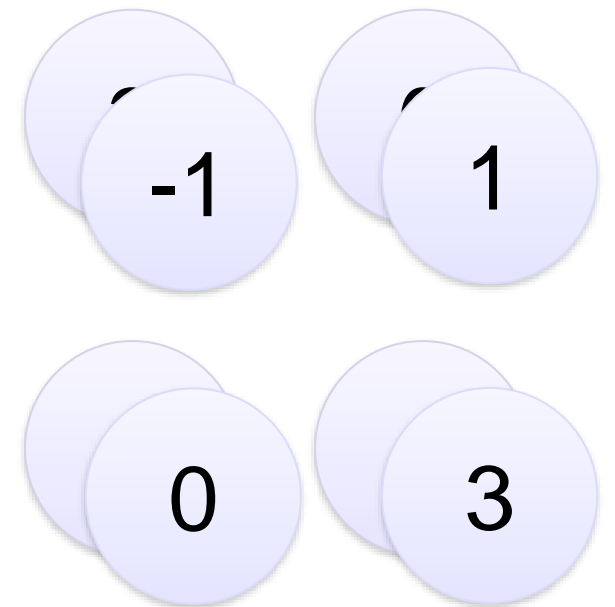
Conv



Max  
Pooling



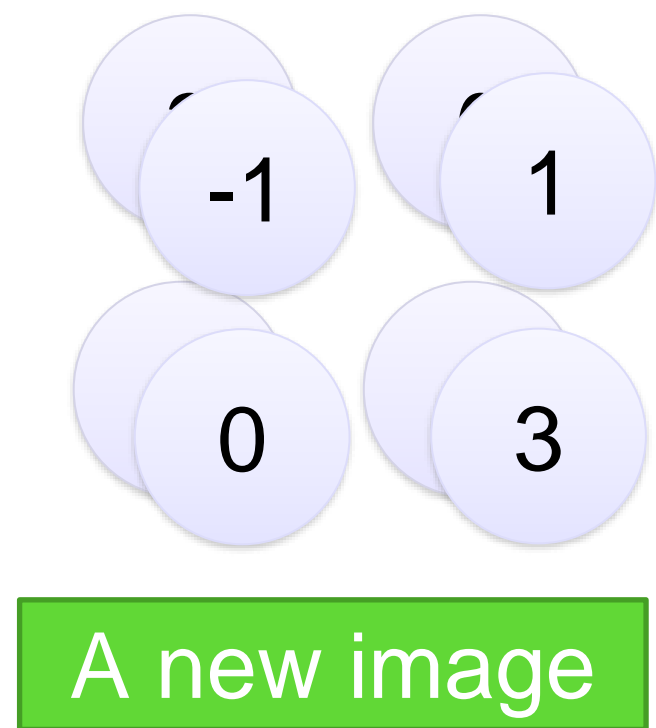
New image  
but smaller



2 x 2 image

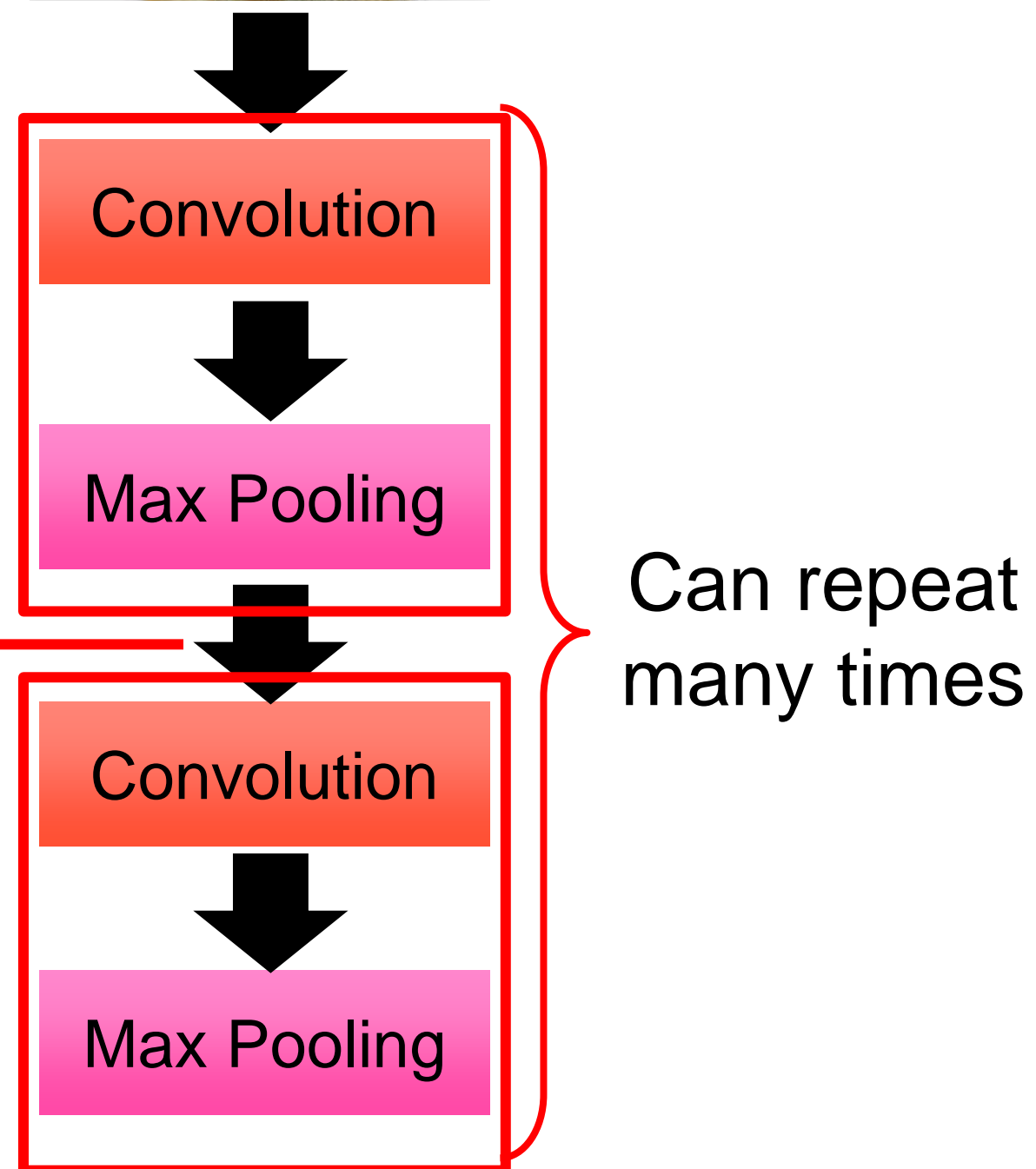
Each filter  
is a channel

# The whole CNN



Smaller than the original image

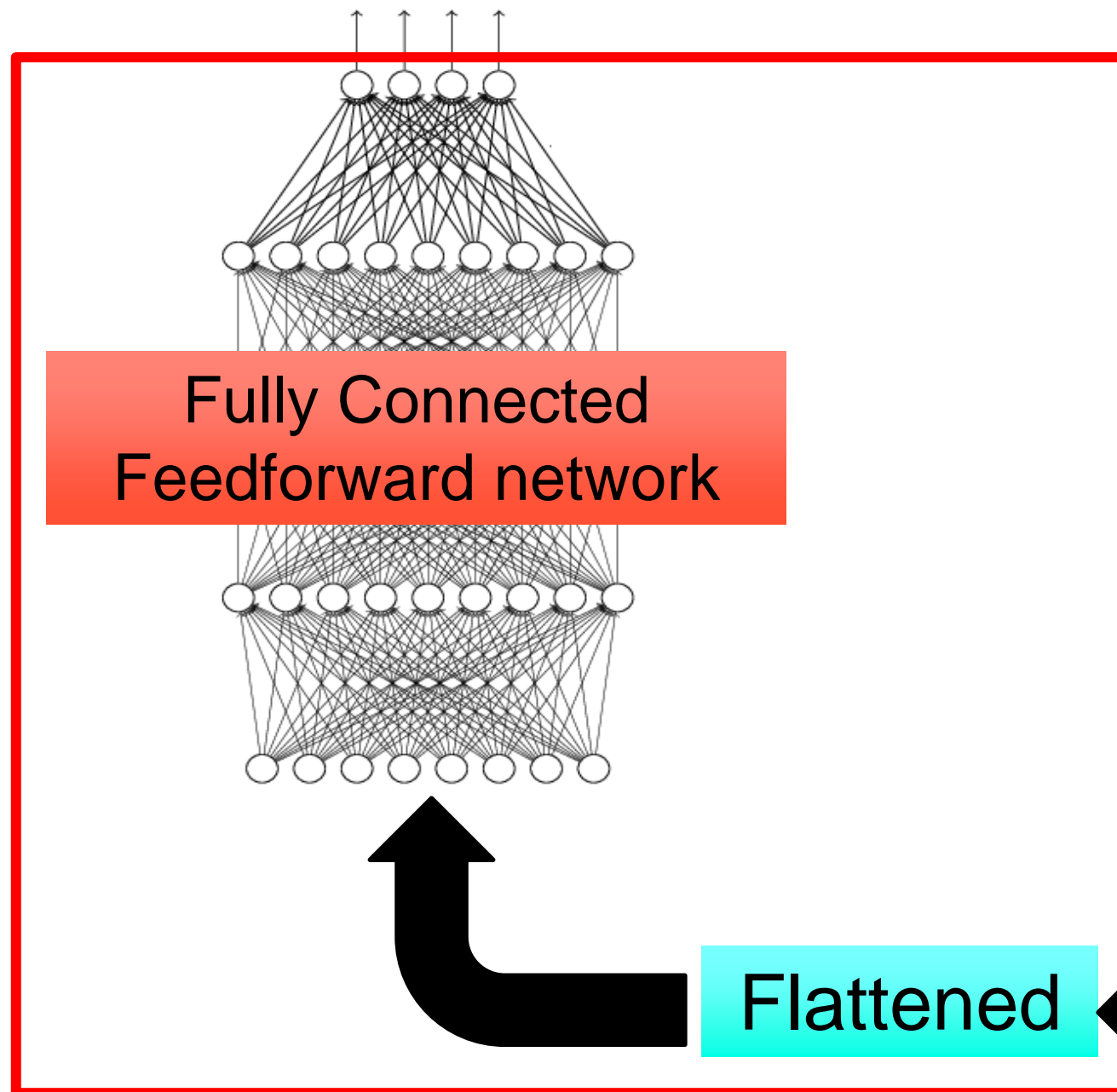
The number of channels is the number of filters





# The whole CNN

cat dog .....



Convolution

Max Pooling

A new image

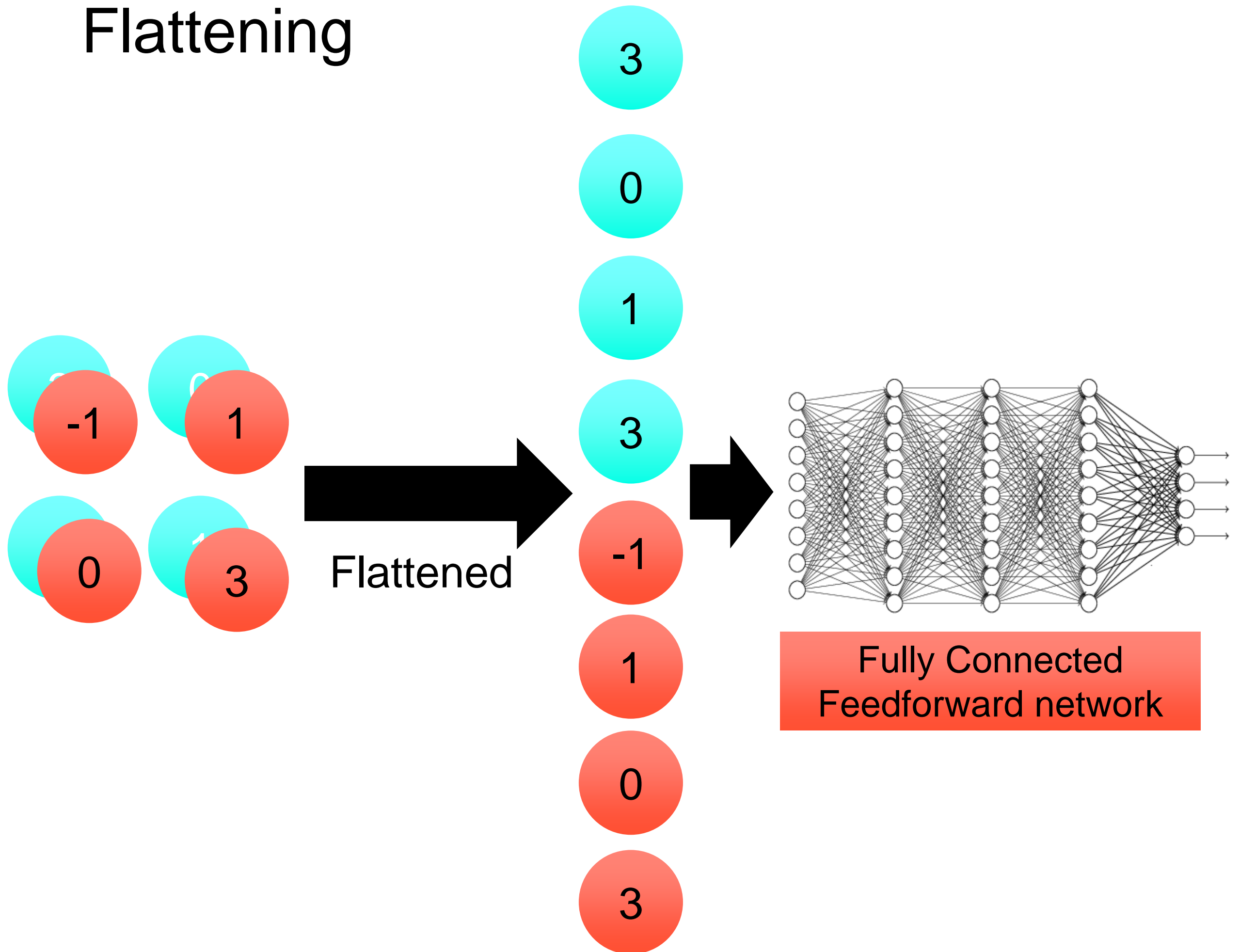
Convolution

Max Pooling

A new image

Flattened

# Flattening



# CNN in Keras

Only modified the *network structure* and *input format* (vector -> 3-D tensor)

```
model2.add( Convolution2D( 25, 3, 3,  
                           input_shape=(28, 28, 1)) )
```

1	-1	-1	1	-1
-1	1	-1	1	-1
-1	-1	-1	1	-1

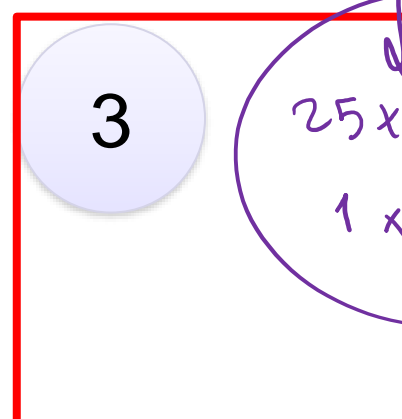
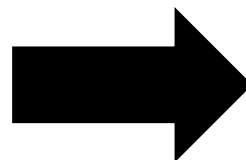
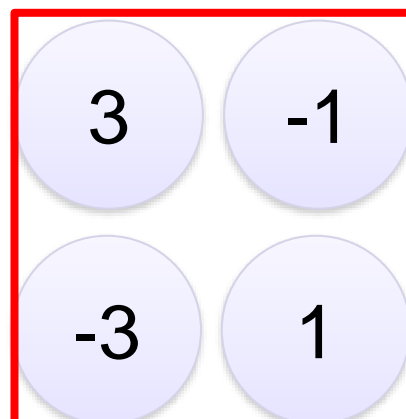
There are  
**25 3x3**  
filters.

Input\_shape = ( 28 , 28 , 1)

28 x 28 pixels

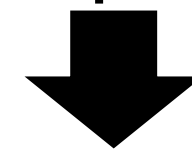
1: black/white, 3: RGB

```
model2.add(MaxPooling2D((2, 2)))
```

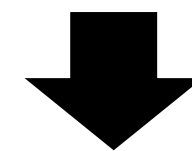


25 x 13 x 13  
1 x 25 x 3 x 3

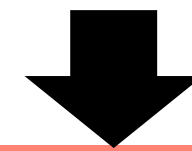
input



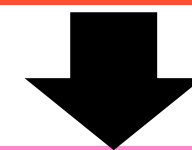
Convolution



Max Pooling



Convolution

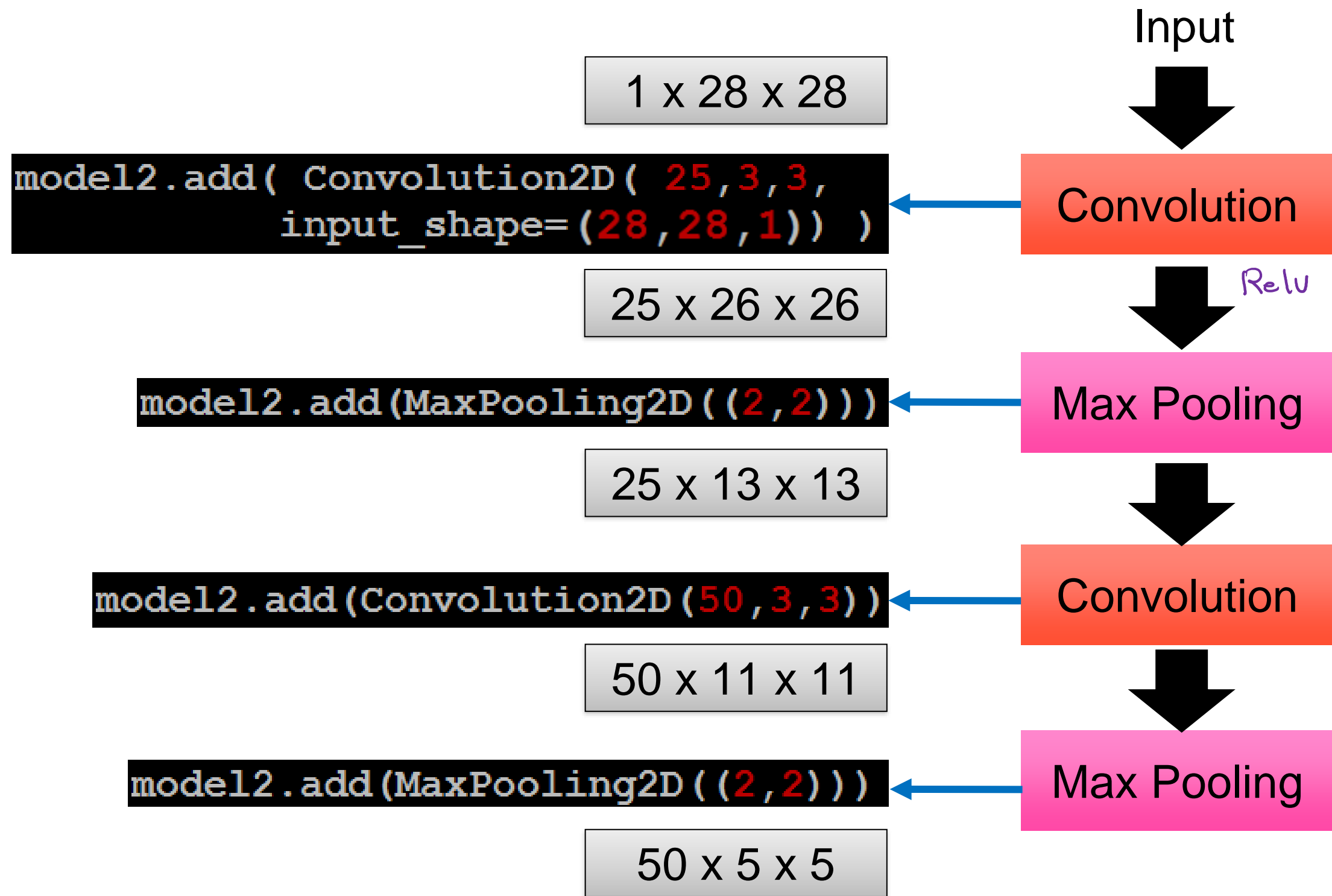


Max Pooling

1 x 1 x 1

# CNN in Keras

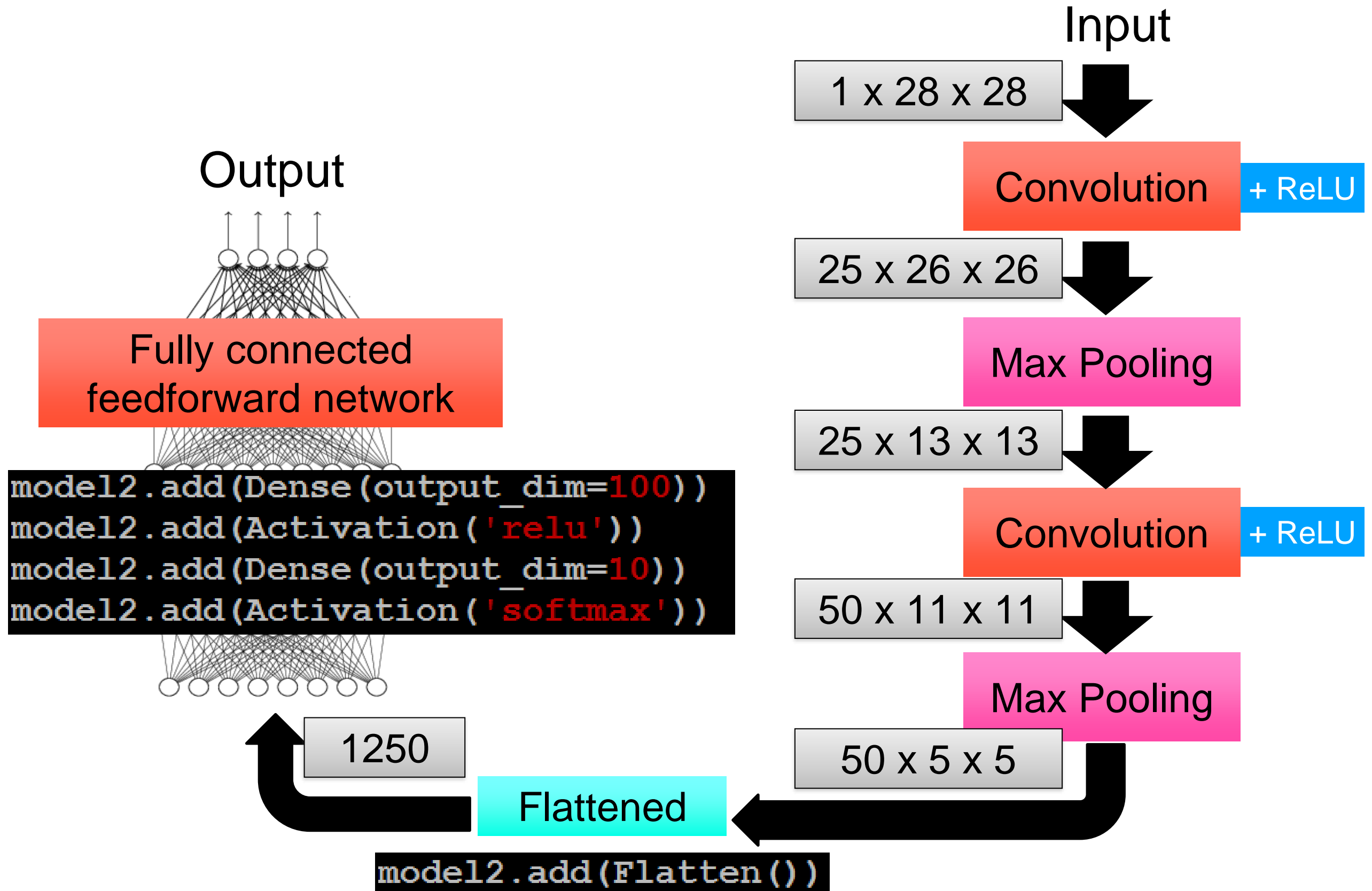
Only modified the *network structure* and *input format* (vector -> 3-D array)



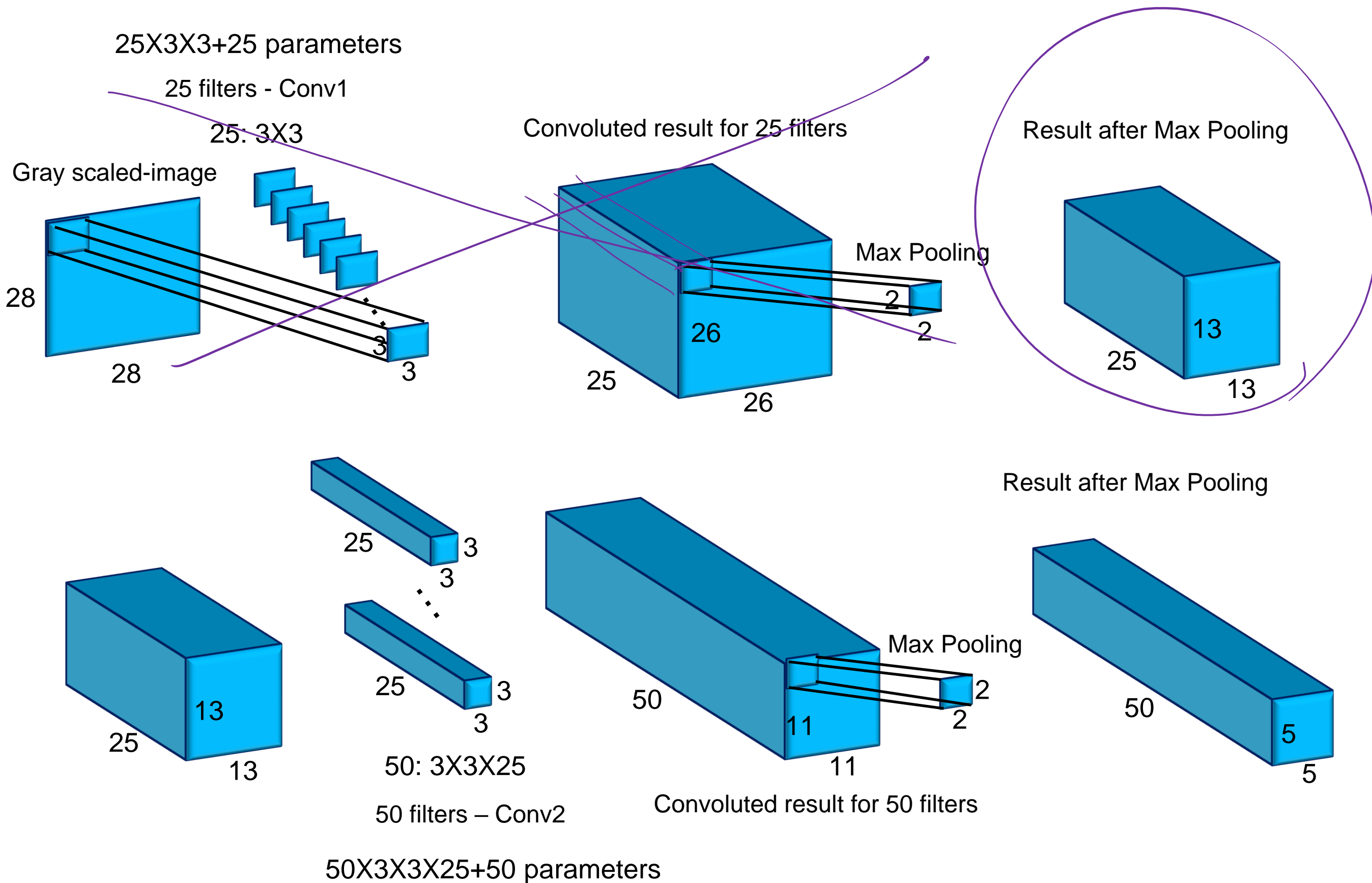


# CNN in Keras

Only modified the *network structure* and *input format* (vector -> 3-D array)

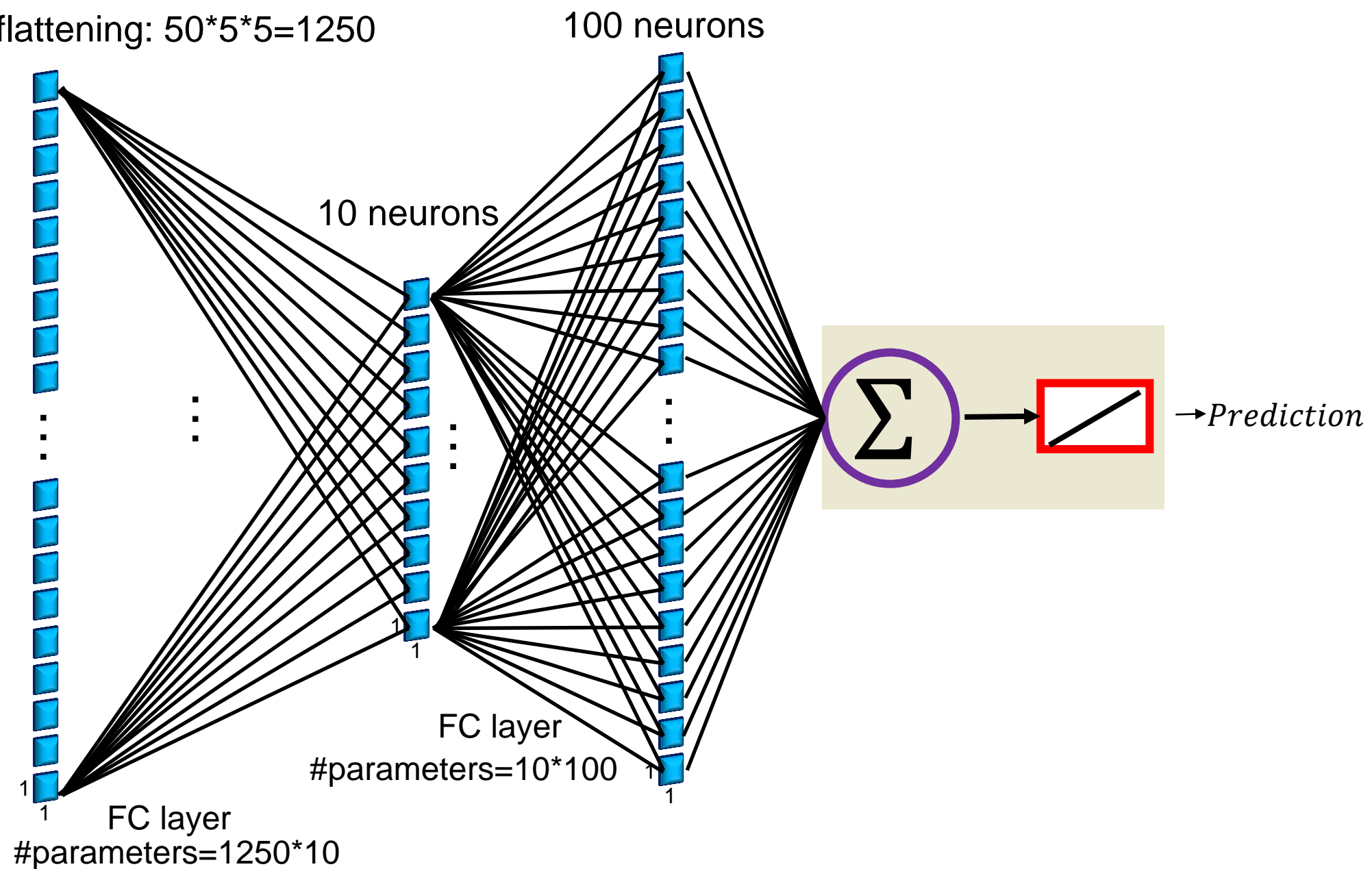
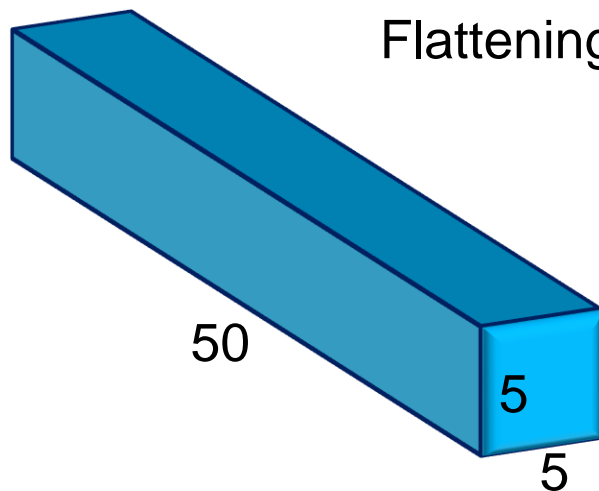


# Number of Parameters



# neurons after flattening:  $50 \times 5 \times 5 = 1250$

Flattening



10 CNN Architecture