



SPPU FOP B.E COMPUTER(2022)

DEEP LEARNING LAB V

Mr.Jameer Kotwal

Asst. Prof.

Pimpri Chinchwad College Of Engg. & Research, Ravet.

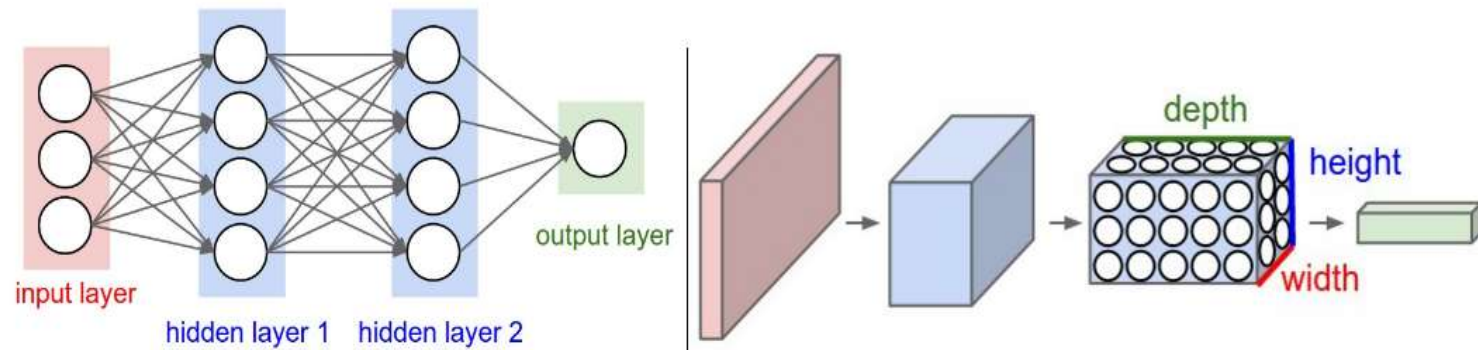


Lab Experiment

Topics:

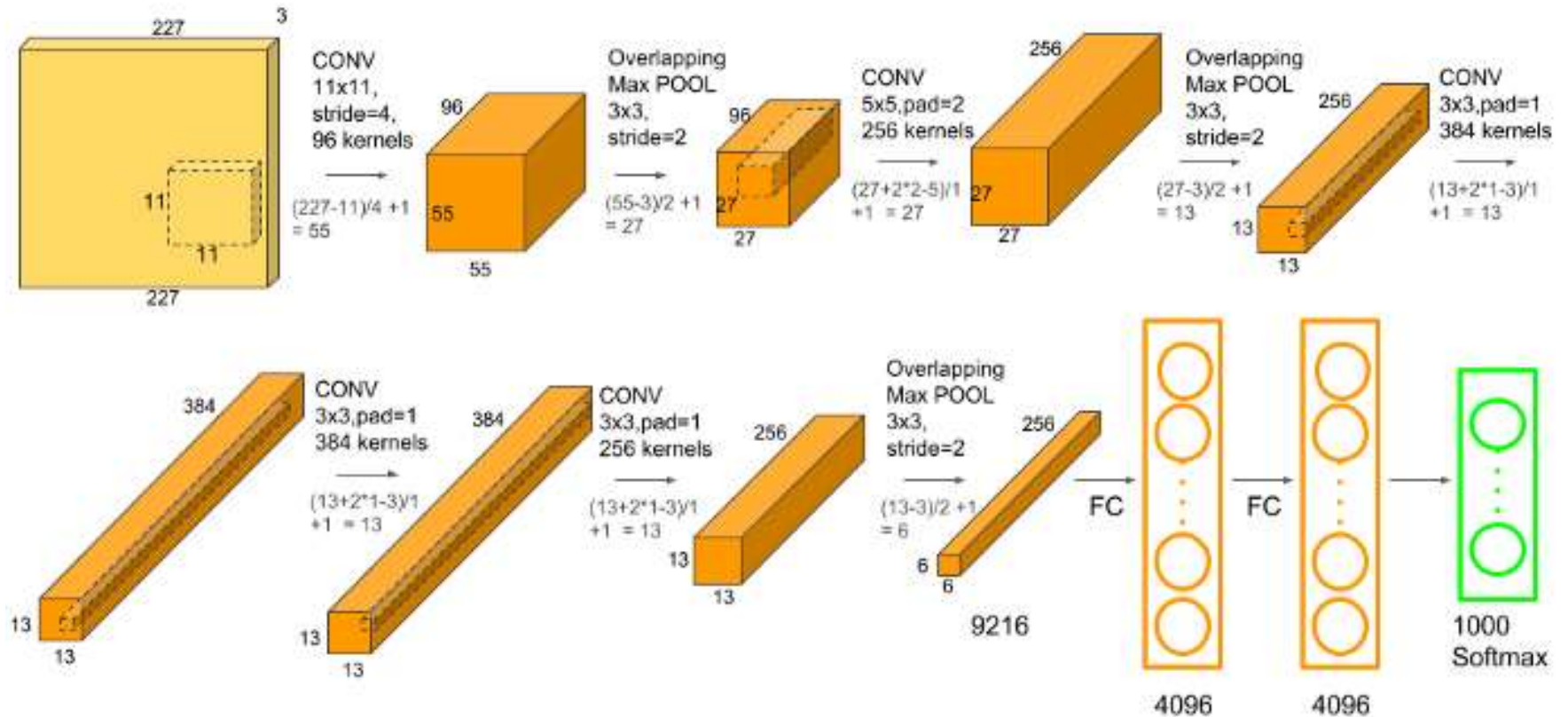
- 1) Linear regression by using Deep Neural network: Implement Boston housing price prediction problem by Linear regression using Deep Neural network. Use Boston House price prediction dataset.
- 2) Classification using Deep neural network
- 3) Convolutional neural network (CNN)
- 4) Recurrent neural network (RNN)

Dense neural network and Convolutional neural network

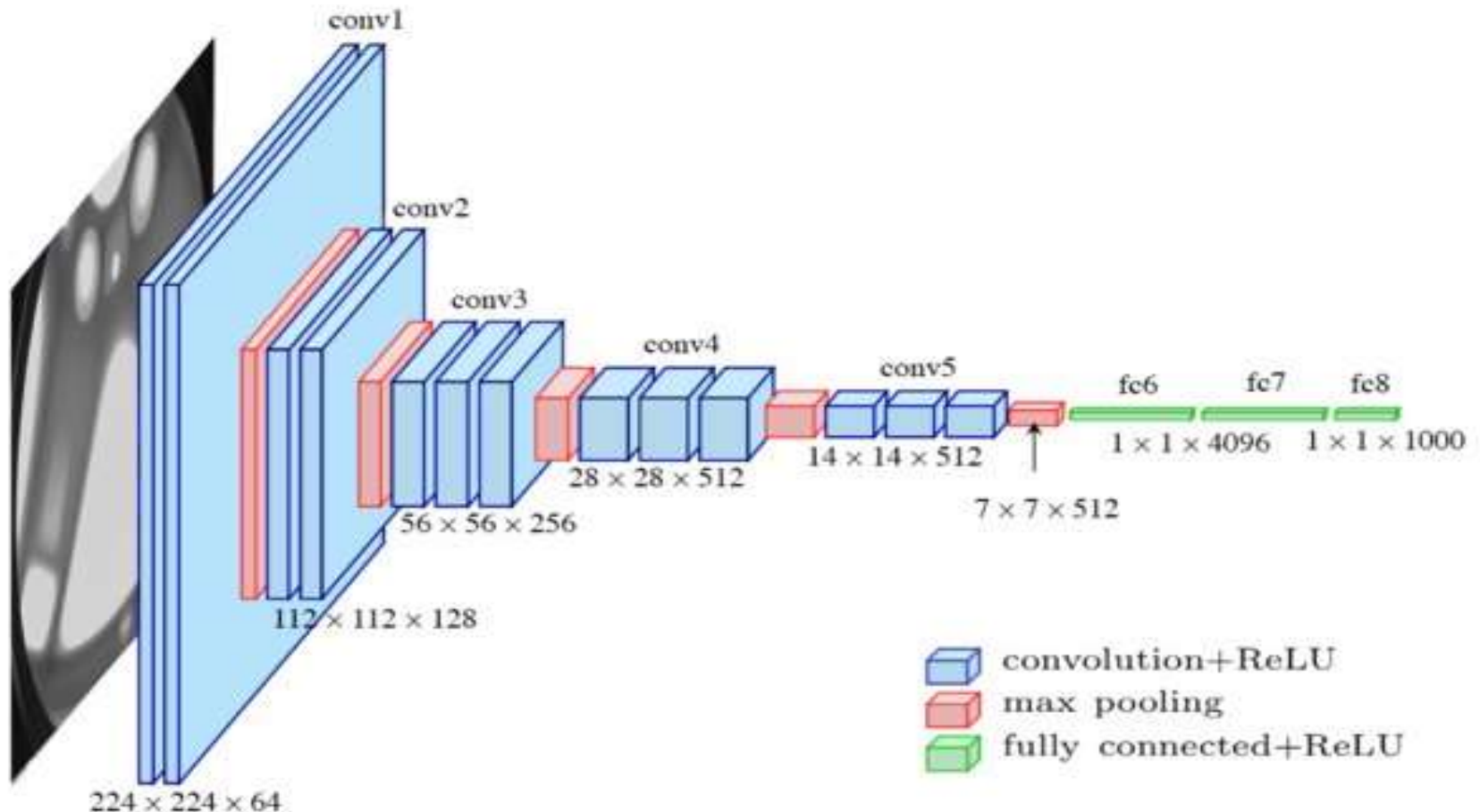


Left: A regular 3-layer Neural Network. Right: A ConvNet arranges its neurons in three dimensions (width, height, depth), as visualized in one of the layers. Every layer of a ConvNet transforms the 3D input volume to a 3D output volume of neuron activations. In this example, the red input layer holds the image, so its width and height would be the dimensions of the image, and the depth would be 3 (Red, Green, Blue channels).

AlexNet Architecture



VGG16 Networks



CIFAR10 dataset and state of the art

The CIFAR-10 dataset consists of 60000 32x32 color images in 10 classes, with 6000 images per class. There are 50000 training images and 10000 test images.

airplane



automobile



bird



cat



deer



dog



frog



horse



ship



truck



Accuracy

| Model | Acc. |
|------------------|--------|
| VGG16 | 92.64% |
| ResNet18 | 93.02% |
| ResNet50 | 93.62% |
| ResNet101 | 93.75% |
| ResNeXt29(32x4d) | 94.73% |
| ResNeXt29(2x64d) | 94.82% |
| DenseNet121 | 95.04% |
| PreActResNet18 | 95.11% |
| DPN92 | 95.16% |

Consider learning an image:

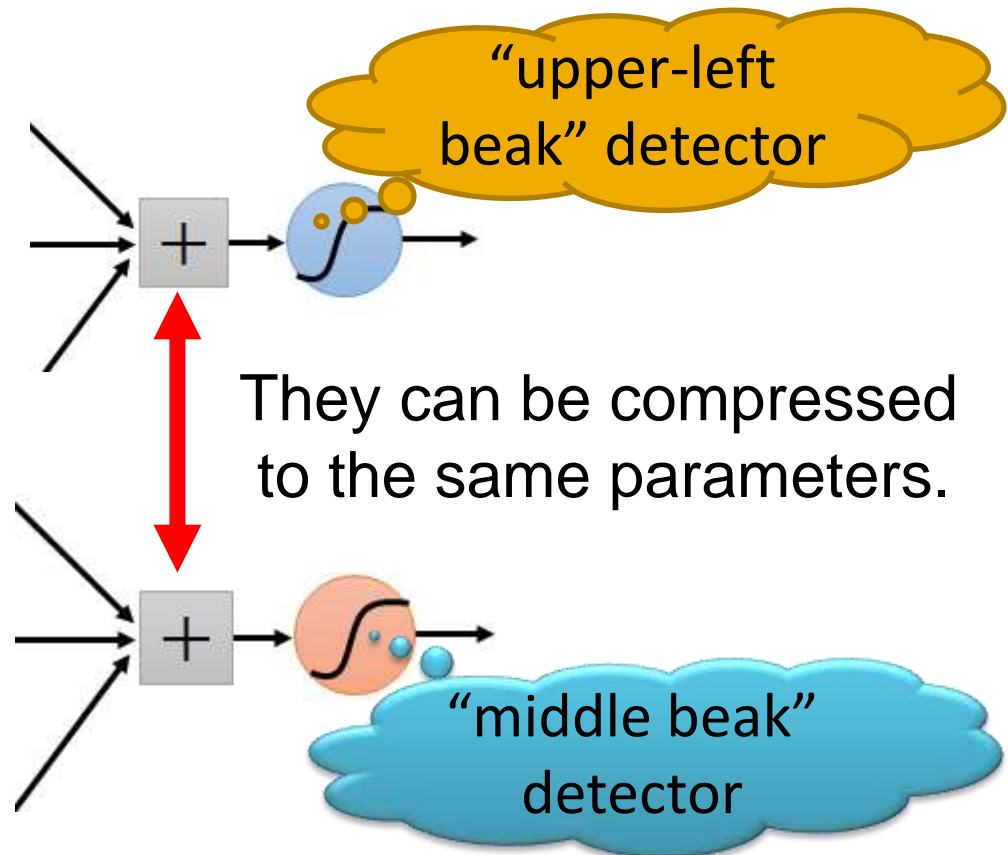
- Some patterns are much smaller than the whole image

Can represent a small region with fewer parameters



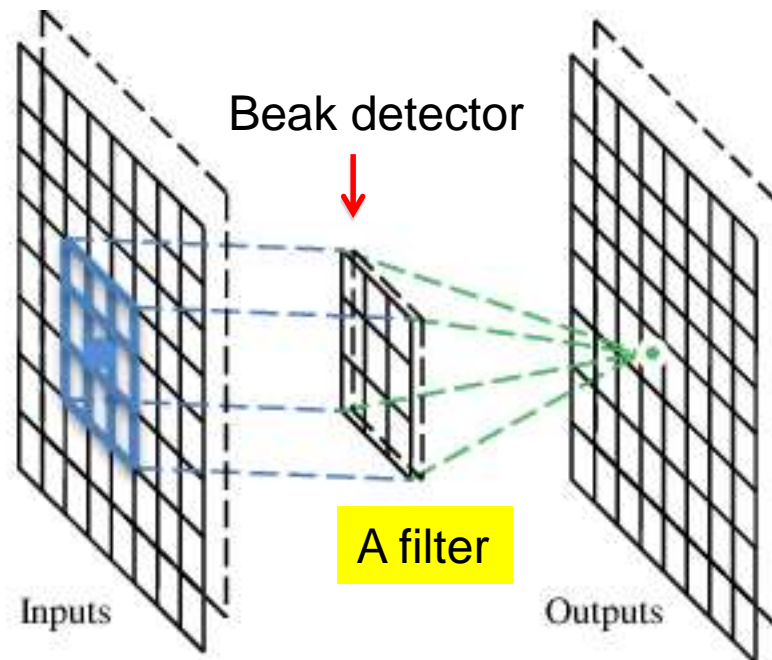
Same pattern appears in different places:
They can be compressed!

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

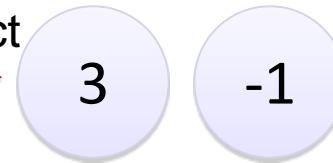
| | | |
|----|----|----|
| 1 | -1 | -1 |
| -1 | 1 | -1 |
| -1 | -1 | 1 |

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 |

Dot
product



6 x 6 image

Convolution

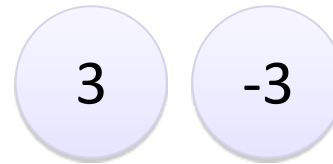
| | | |
|----|----|----|
| 1 | -1 | -1 |
| -1 | 1 | -1 |
| -1 | -1 | 1 |

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

| | | |
|----|----|----|
| 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

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

Convolutio

| | | |
|----|---|----|
| -1 | 1 | -1 |
| -1 | 1 | -1 |
| -1 | 1 | -1 |

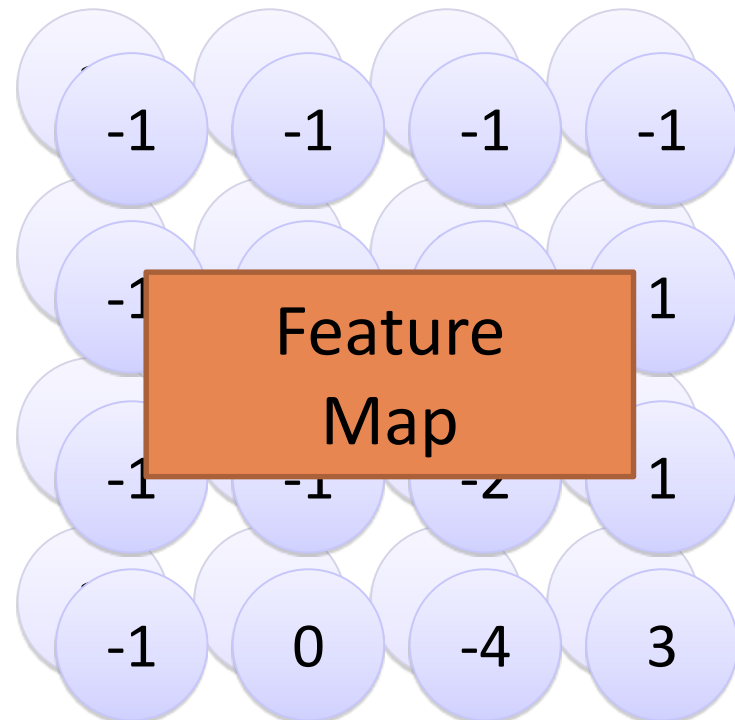
Filter 2

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 |

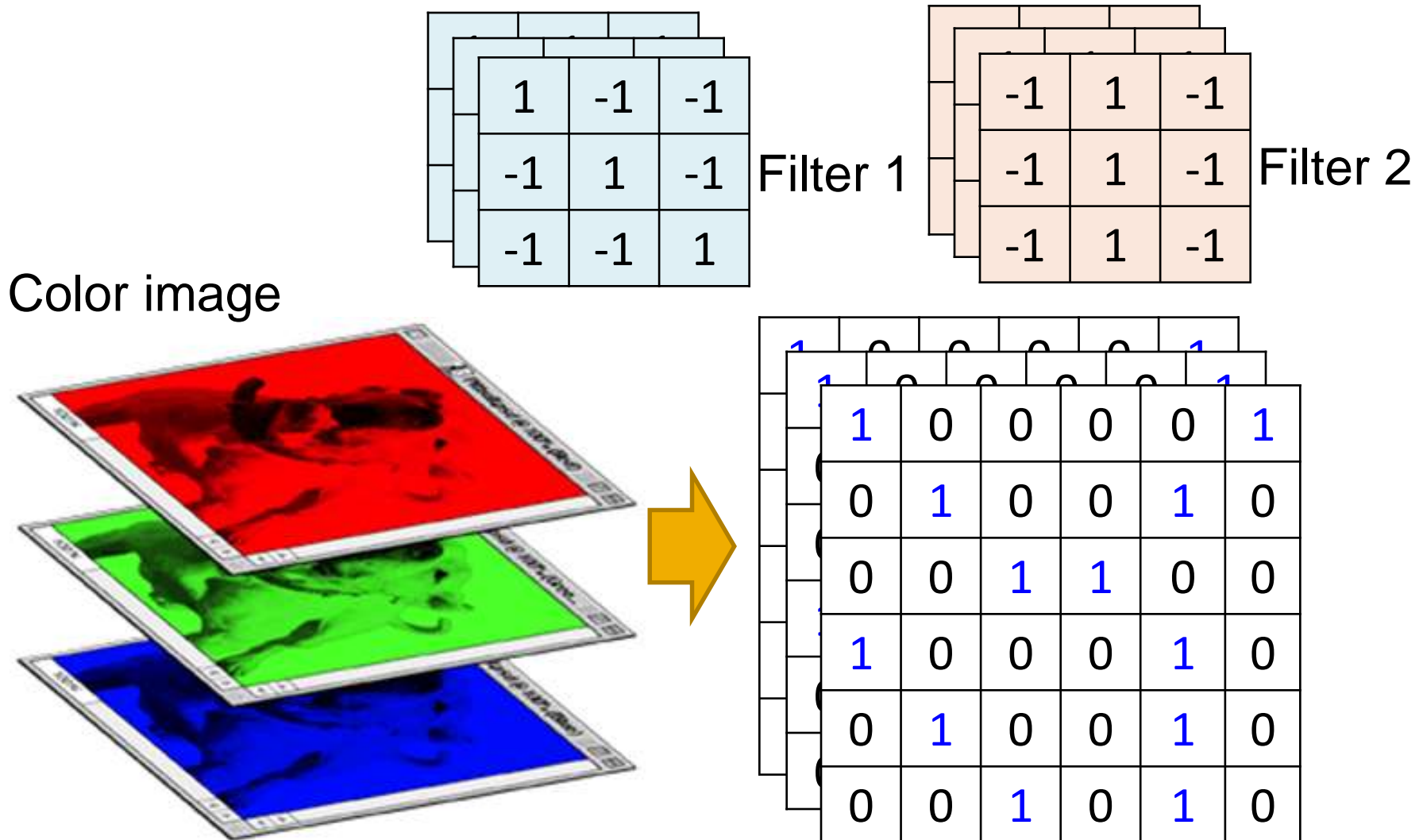
6 x 6 image

Repeat this for each filter

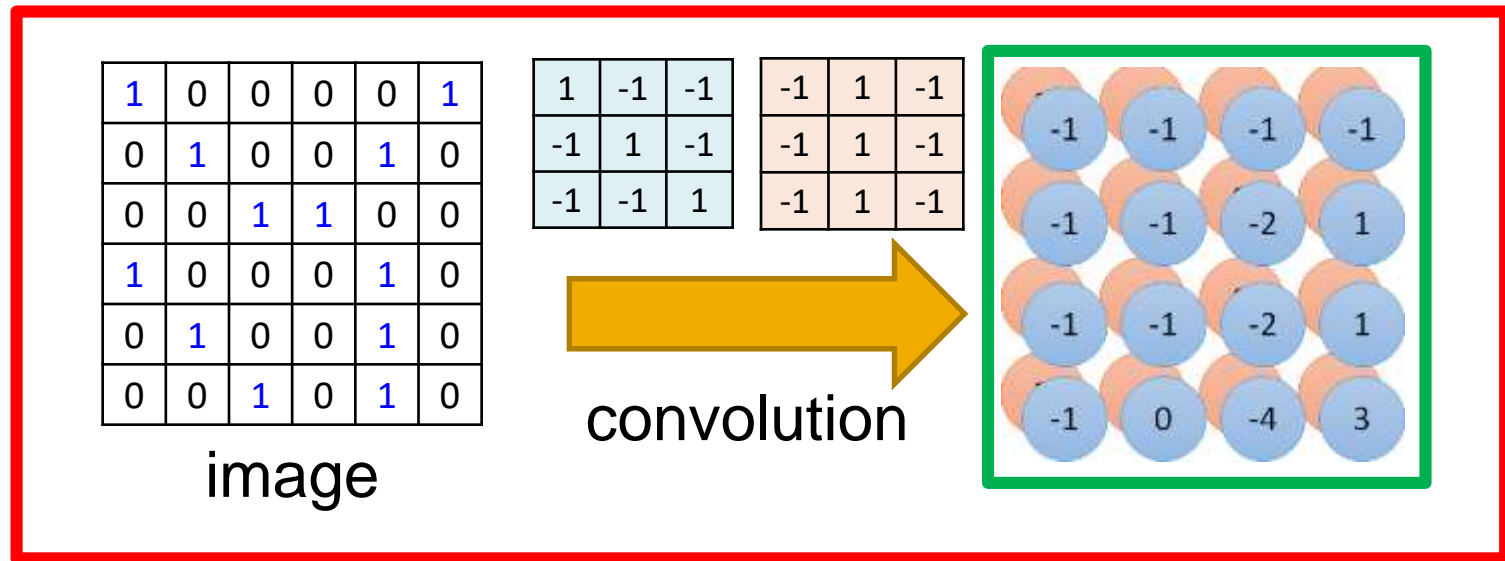


Two 4 x 4 images
Forming 2 x 4 x 4 matrix

Color image: RGB 3 channels

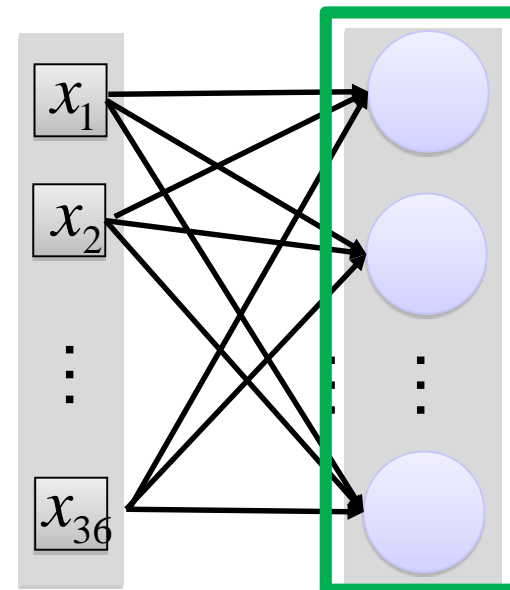


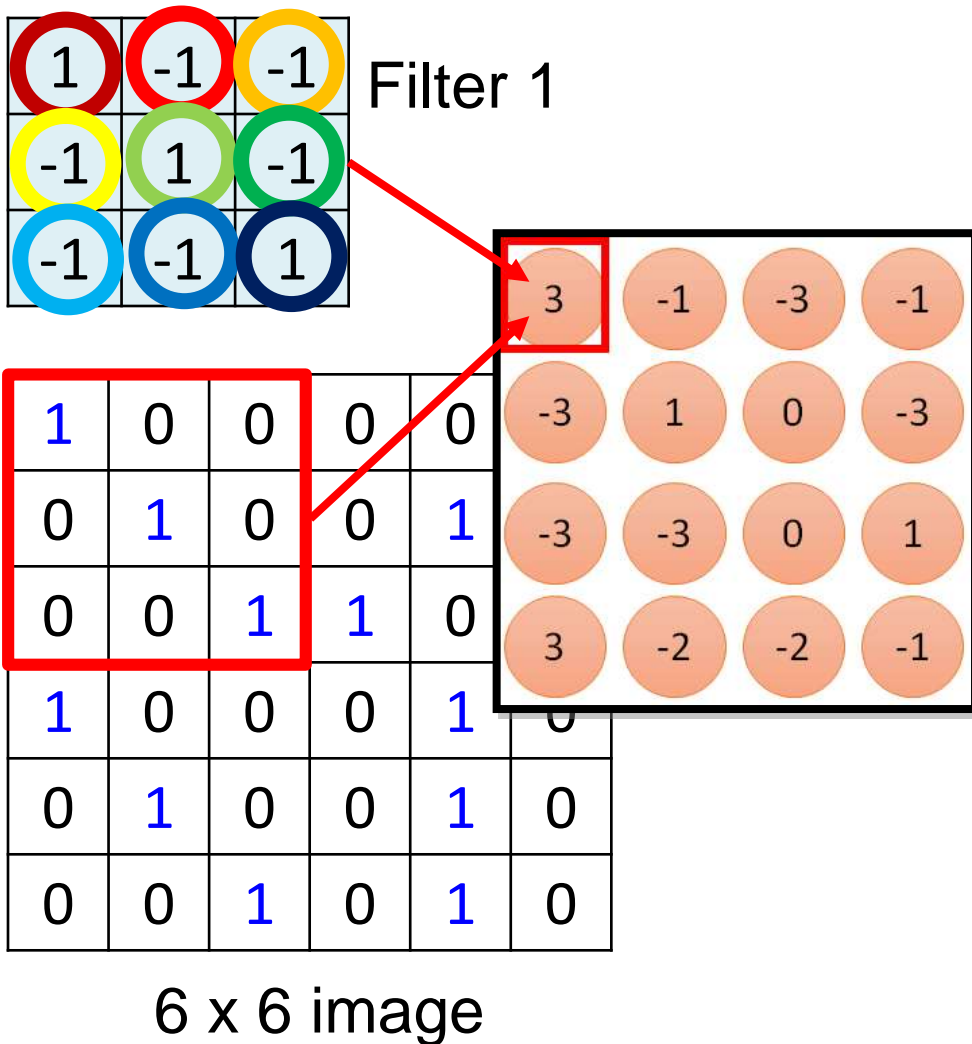
Convolution v.s. Fully Connected



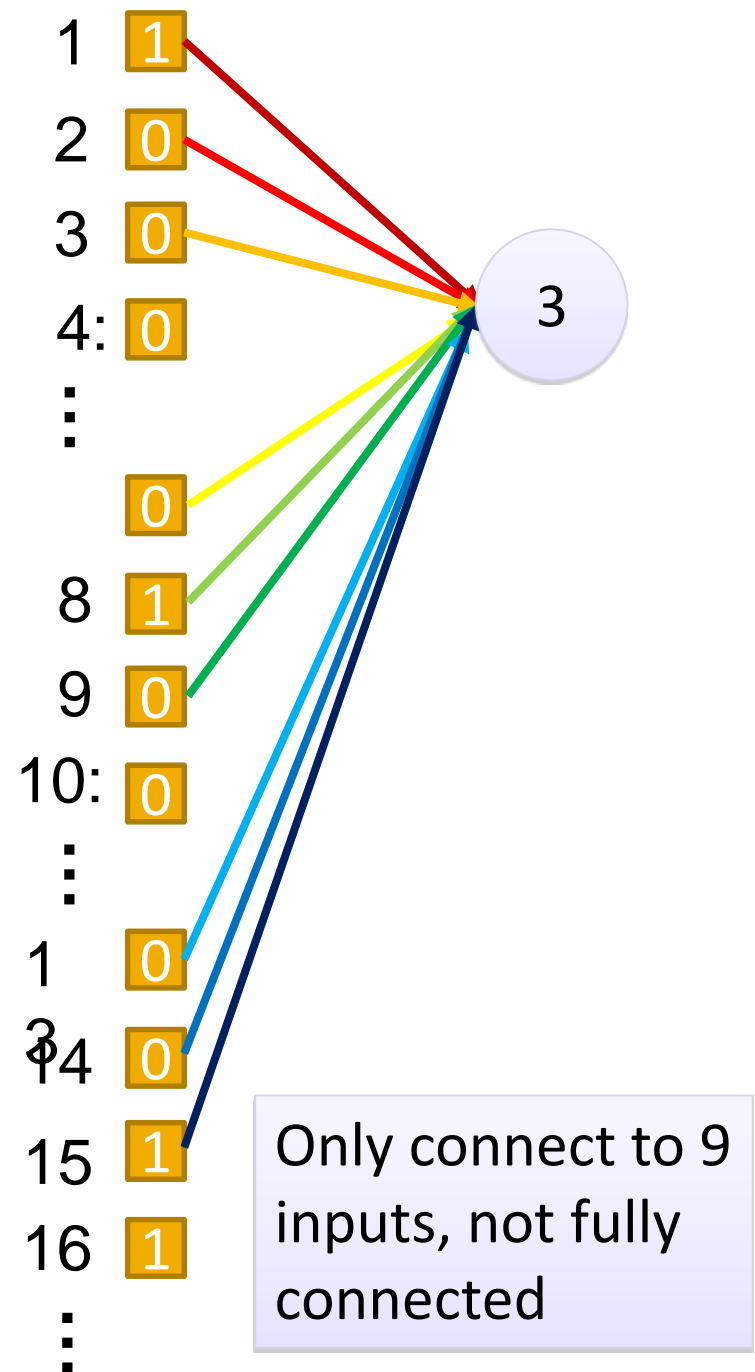
Fully-
connected

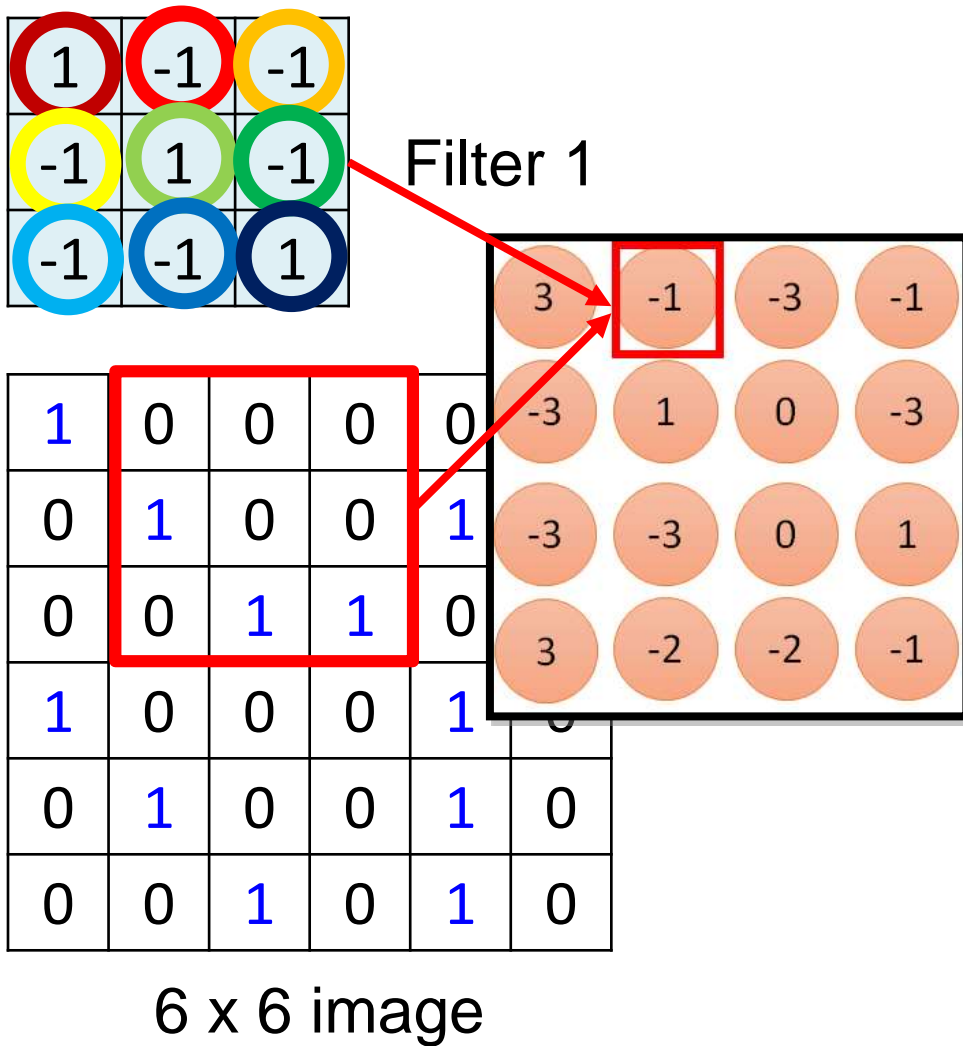
| | | | | | |
|---|---|---|---|---|---|
| 1 | 0 | 0 | 0 | 0 | 1 |
| 0 | 1 | 0 | 0 | 1 | 0 |
| 0 | 0 | 1 | 1 | 0 | 0 |
| 1 | 0 | 0 | 0 | 1 | 0 |
| 0 | 1 | 0 | 0 | 1 | 0 |
| 0 | 0 | 1 | 0 | 1 | 0 |





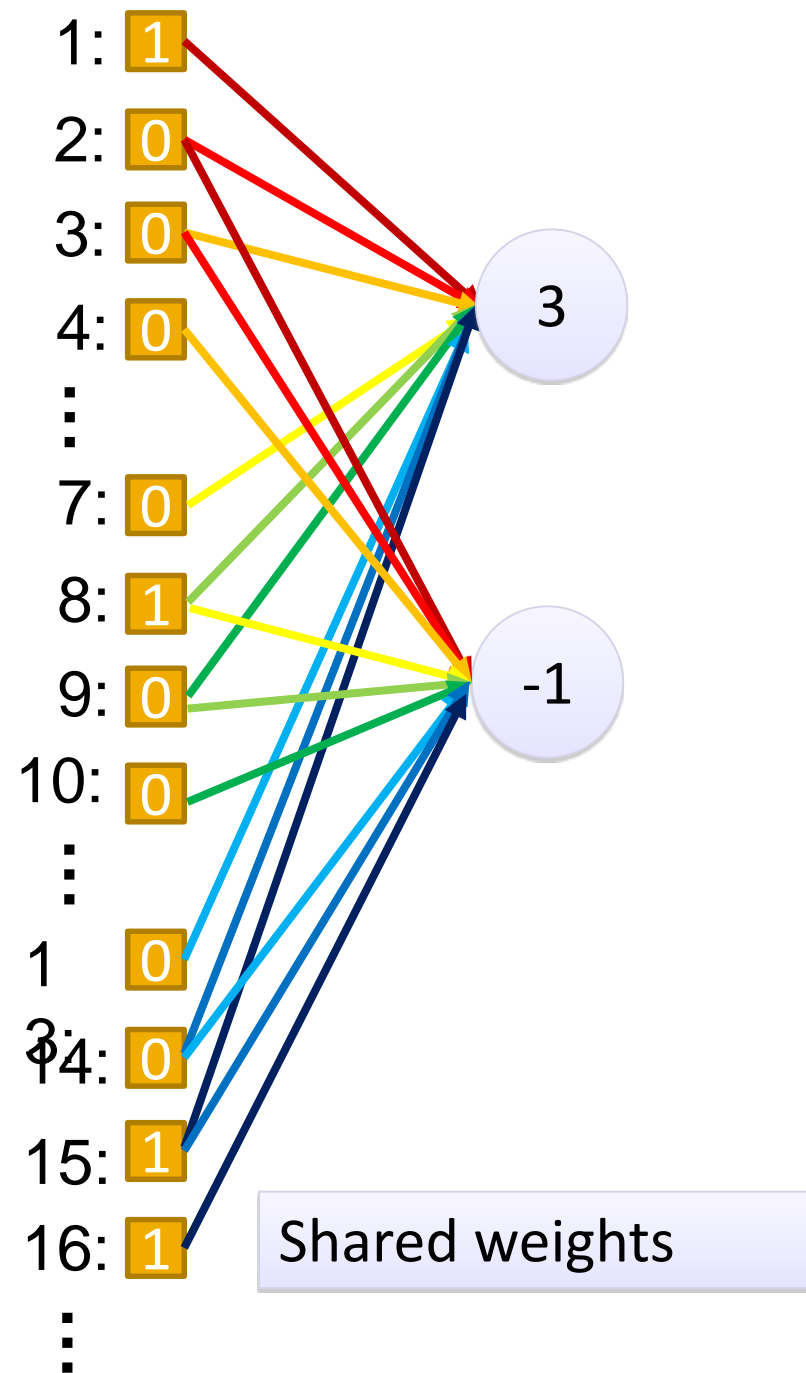
fewer parameters!





Fewer parameters

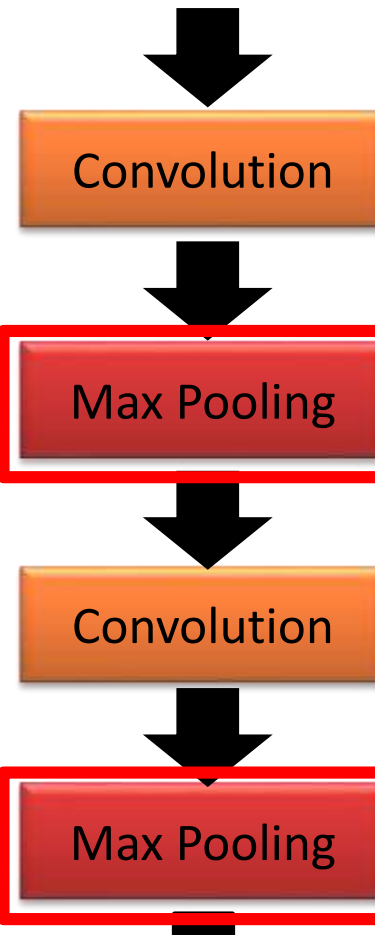
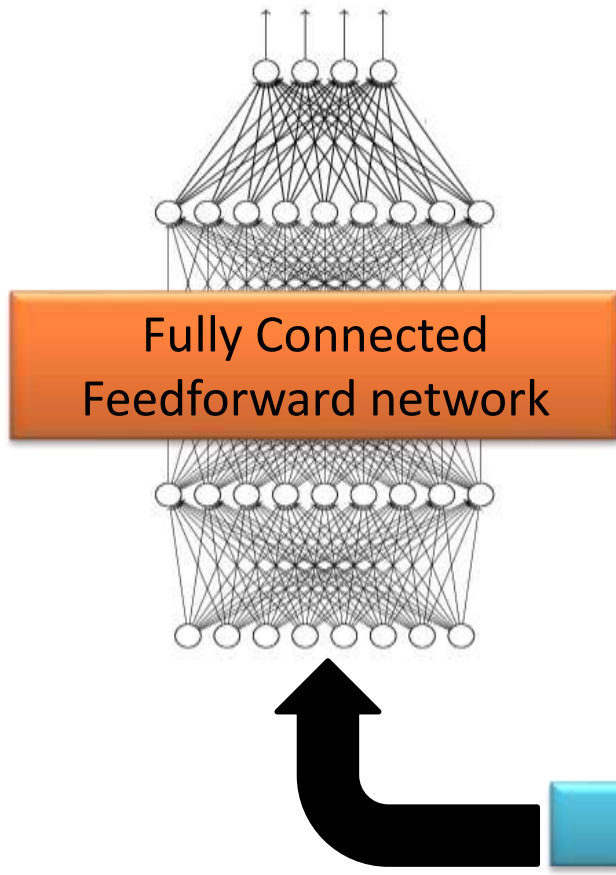
Even fewer parameters



The whole



cat dog



Flattened

Max Pooling

| | | |
|----|----|----|
| 1 | -1 | -1 |
| -1 | 1 | -1 |
| -1 | -1 | 1 |

Filter 1

| | | |
|----|---|----|
| -1 | 1 | -1 |
| -1 | 1 | -1 |
| -1 | 1 | -1 |

Filter 2

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

| | | | |
|----|----|----|----|
| -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



fewer parameters to characterize the image

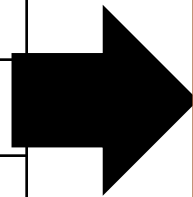
A CNN compresses a fully connected network in two ways:

- 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 |

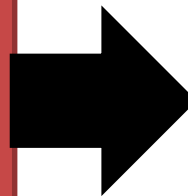
6 x 6 image



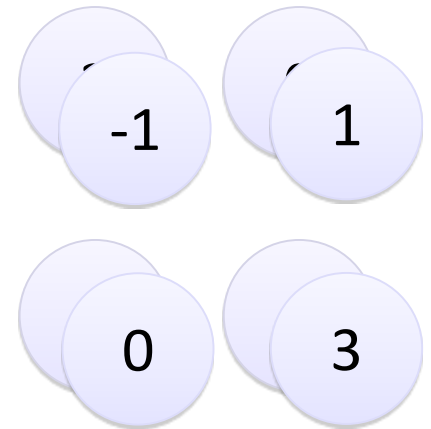
Conv



Max
Pooling



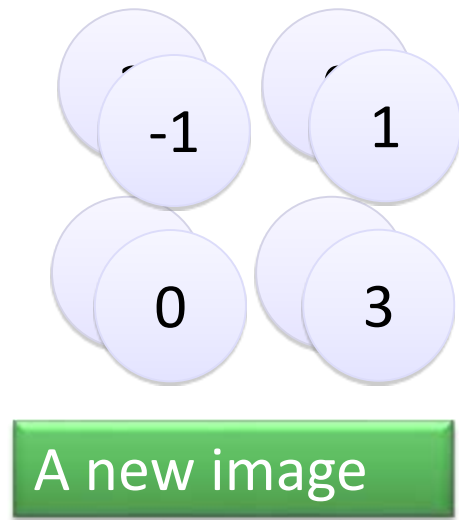
New image
but smaller



2 x 2 image

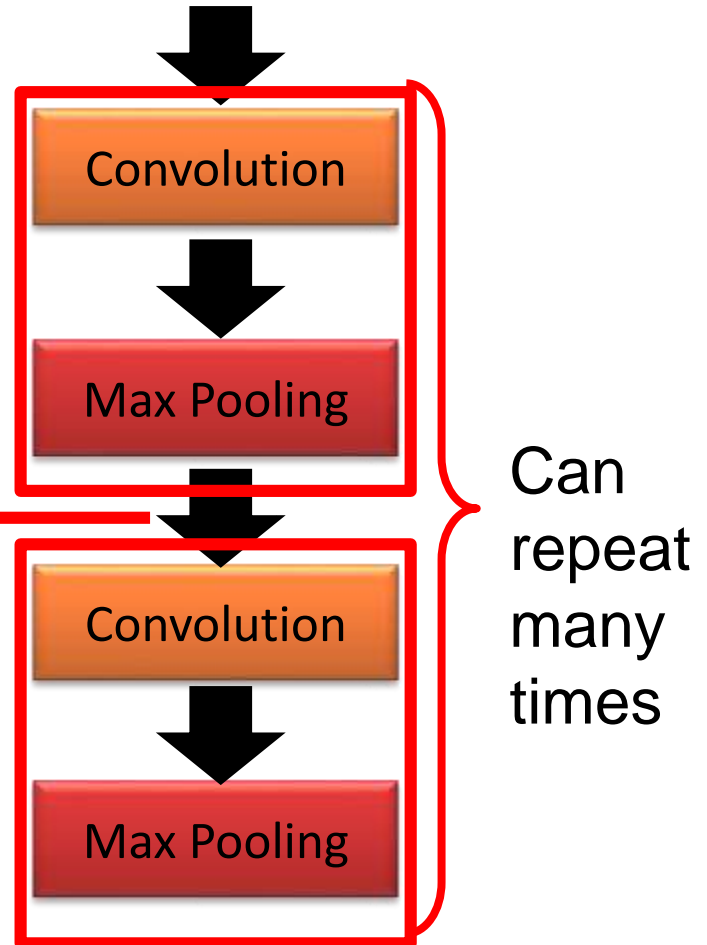
Each filter
is a channel

The whole



Smaller than the original image

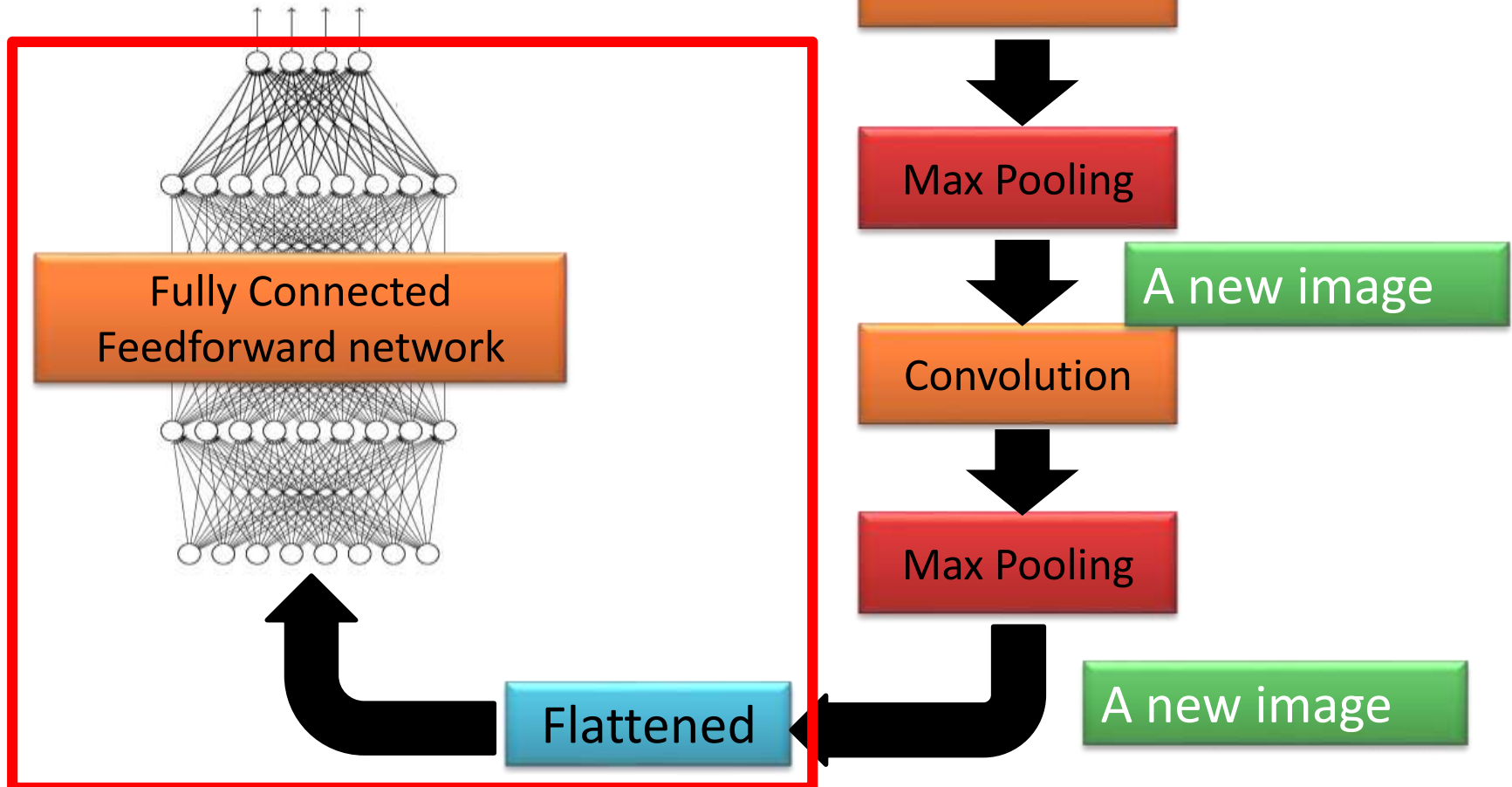
The number of channels is the number of filters

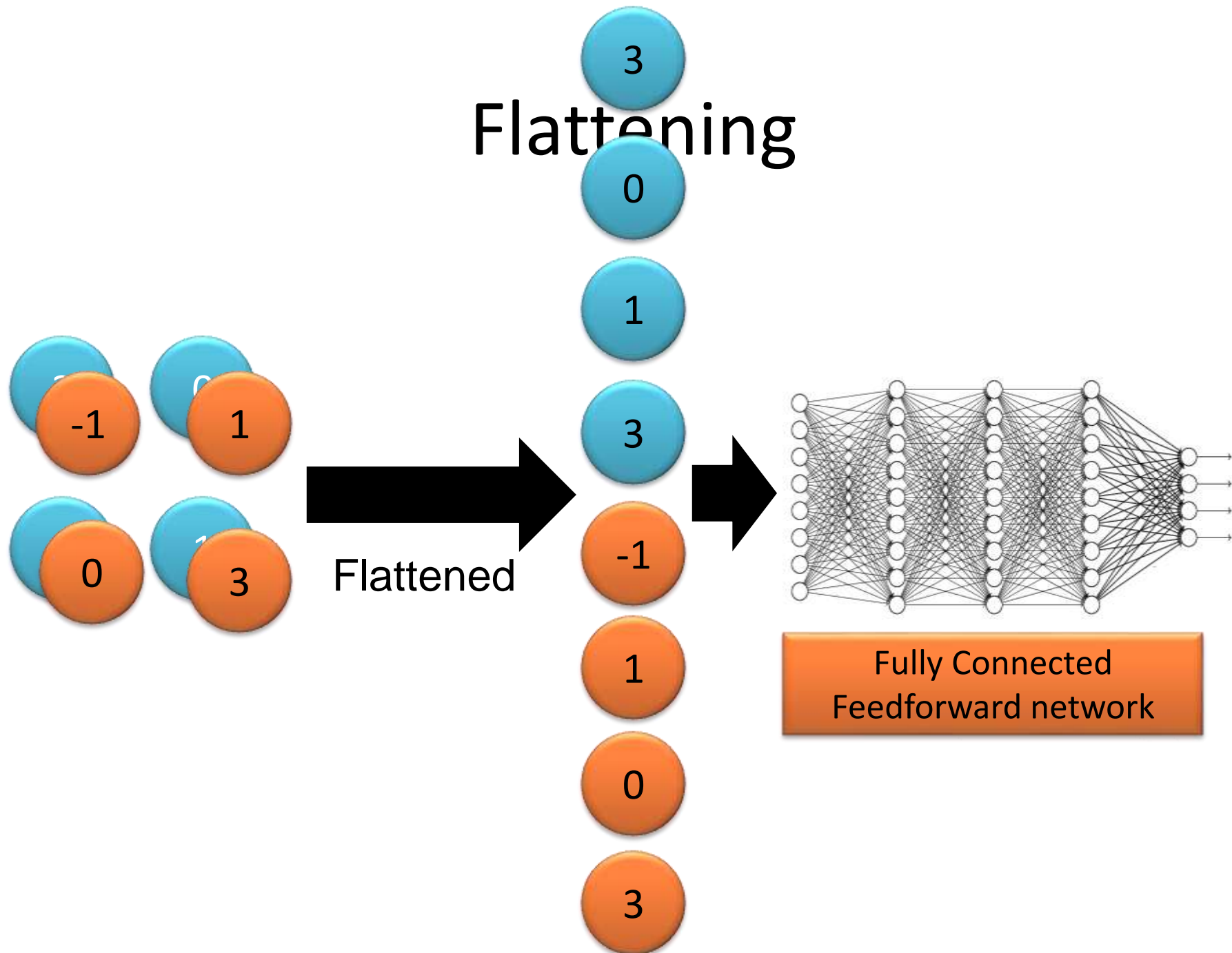


The whole



cat dog





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 |

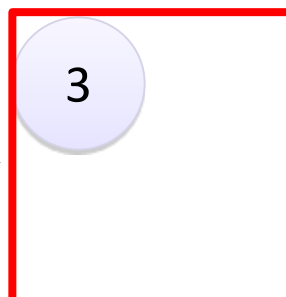
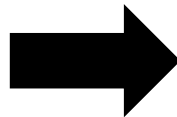
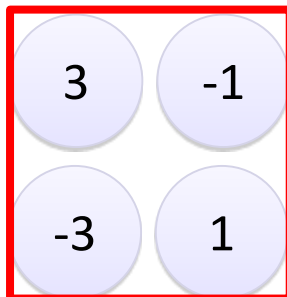
There are
25 3x3
filters.

Input_shape = (28 , 28 , 1)

28 x 28 pixels

1: black/white, 3: RGB

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



input

Convolution

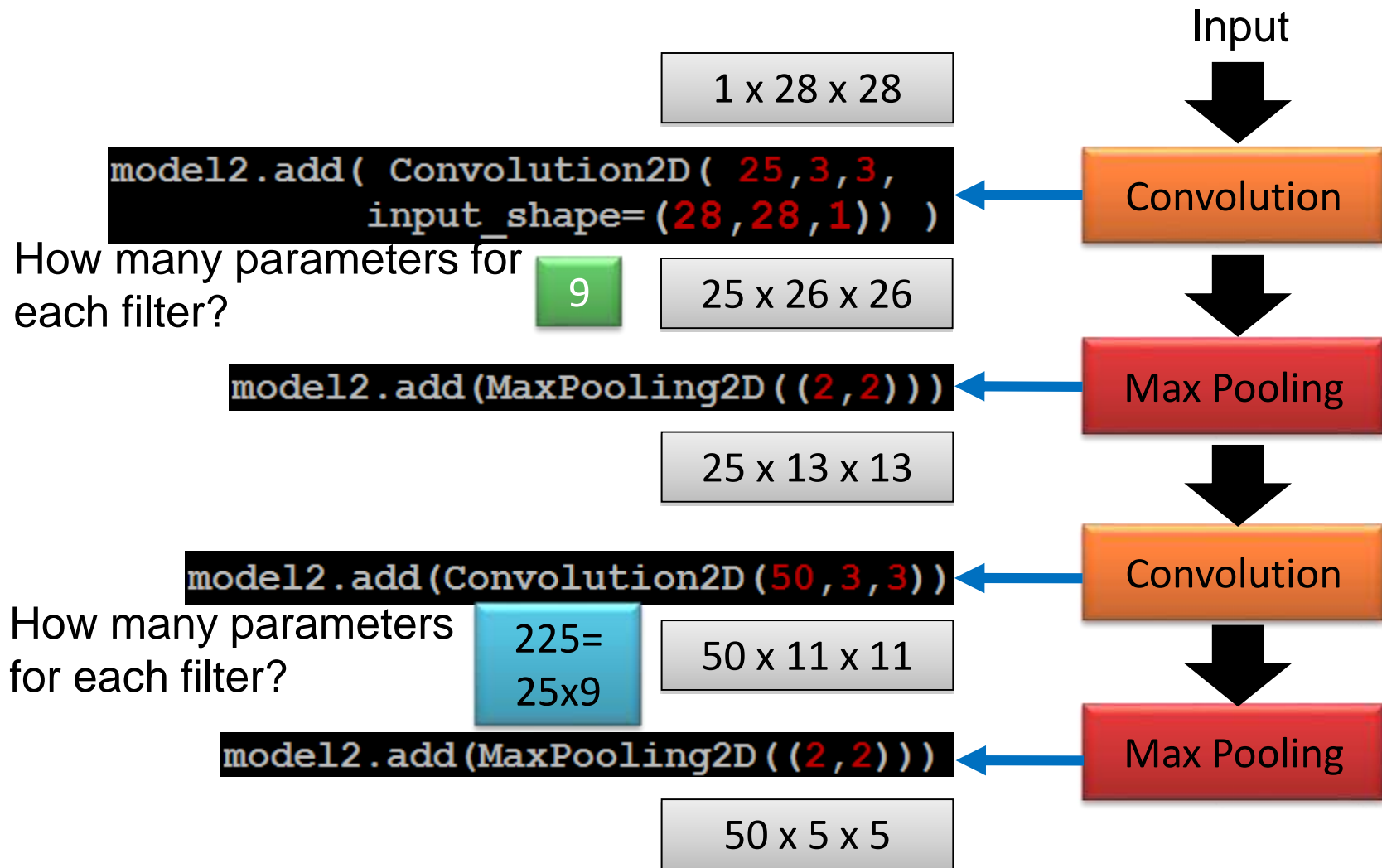
Max Pooling

Convolution

Max Pooling

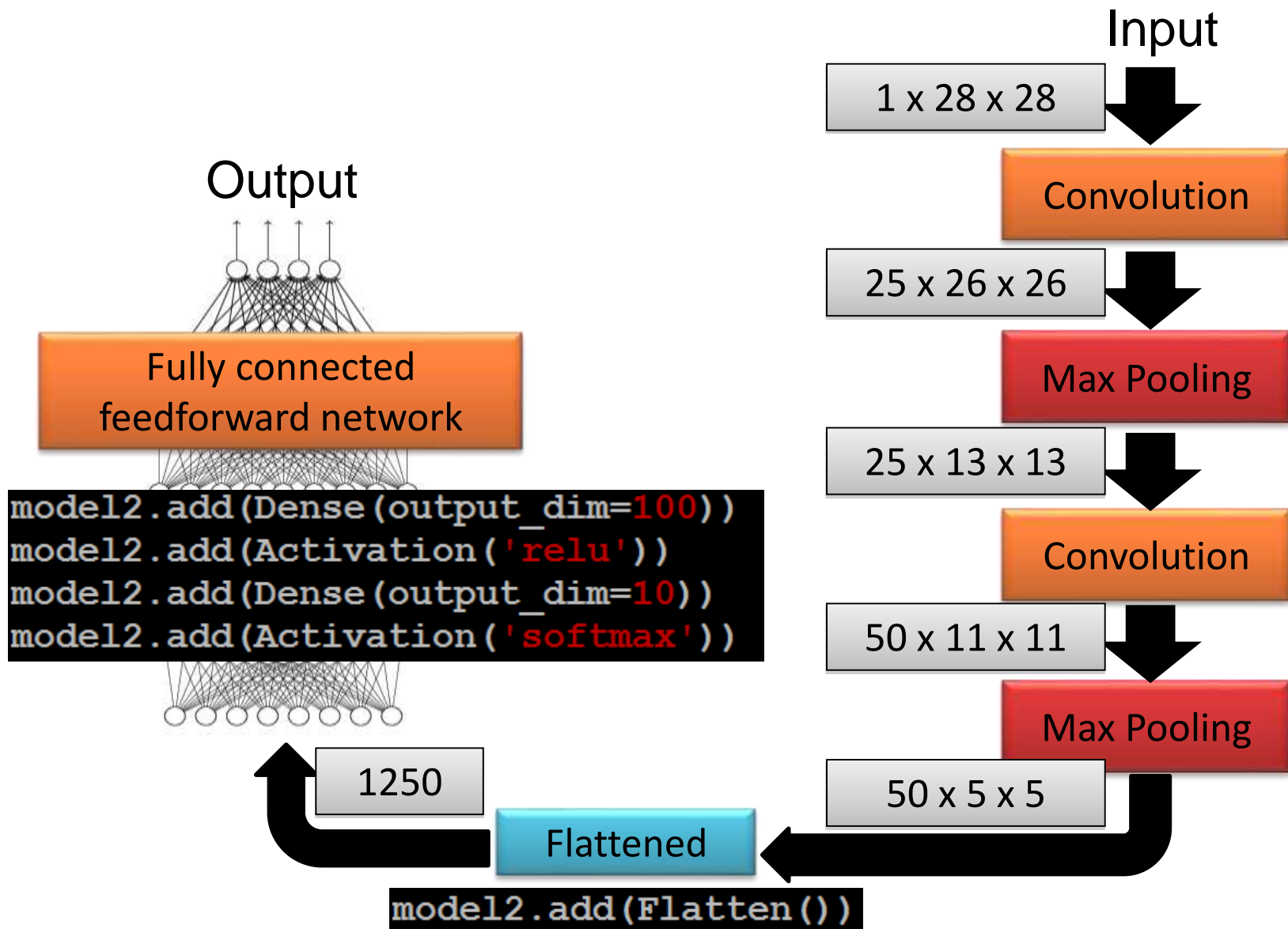
CNN in Keras

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)*



Handwritten Digit

| | | | | |
|----|----|----|----|----|
| -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | 1 | -1 |
| -1 | 1 | 1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | 1 | -1 | -1 |
| -1 | 1 | -1 | -1 | -1 |

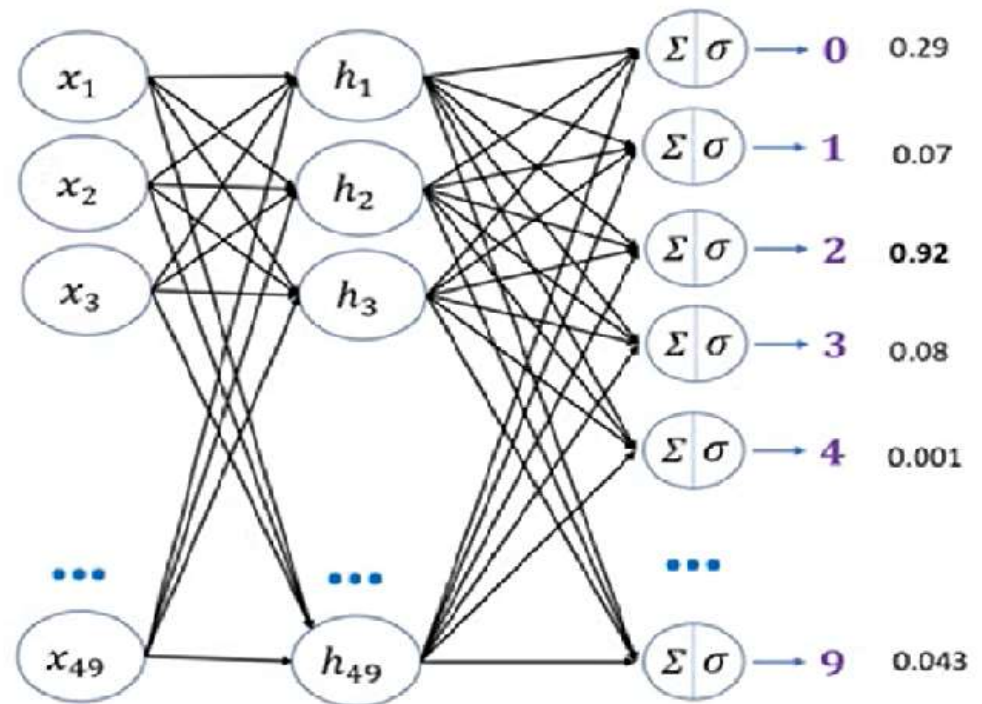
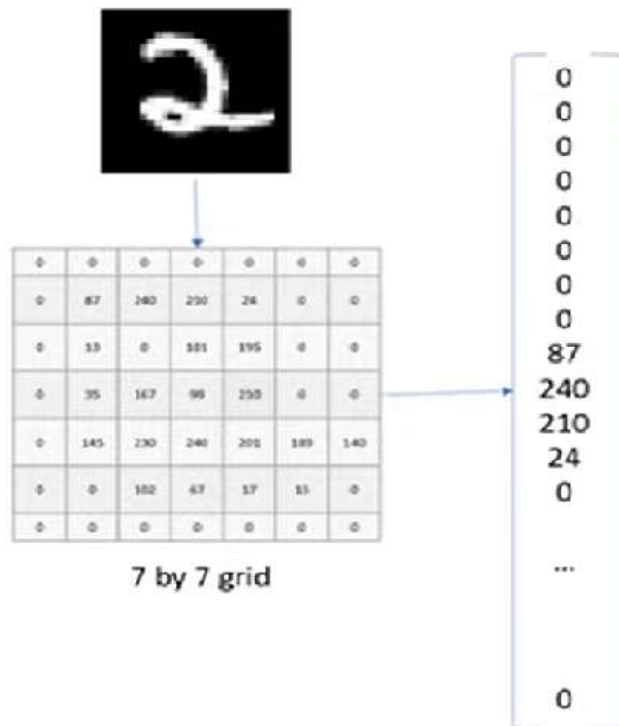
| | | | | |
|----|----|----|----|----|
| -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | 1 | -1 |
| -1 | 1 | 1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | 1 | -1 | -1 |
| -1 | 1 | -1 | -1 | -1 |

Handwritten Digit

Location Shifted



Handwritten Digit



$$49 \times 49 = 2401$$

$$49 \times 10 = 490$$

Handwritten Digit



Image size = $1920 \times 1080 \times 3$

First layer neurons = $1920 \times 1080 \times 3 \sim 6 \text{ million}$

Hidden layer neurons = Let's say you keep it $\sim 4 \text{ million}$

Weights between input and hidden layer = $6 \text{ mil} \times 4 \text{ mil}$
 $= 24 \text{ million}$

Handwritten Digit



Koala's **eye**? = Y



Koala's **nose**? = Y



Koala's **ears**? = Y



Koala's **hands**? = Y



Koala's **legs**? = Y



Koala's **head**? = Y

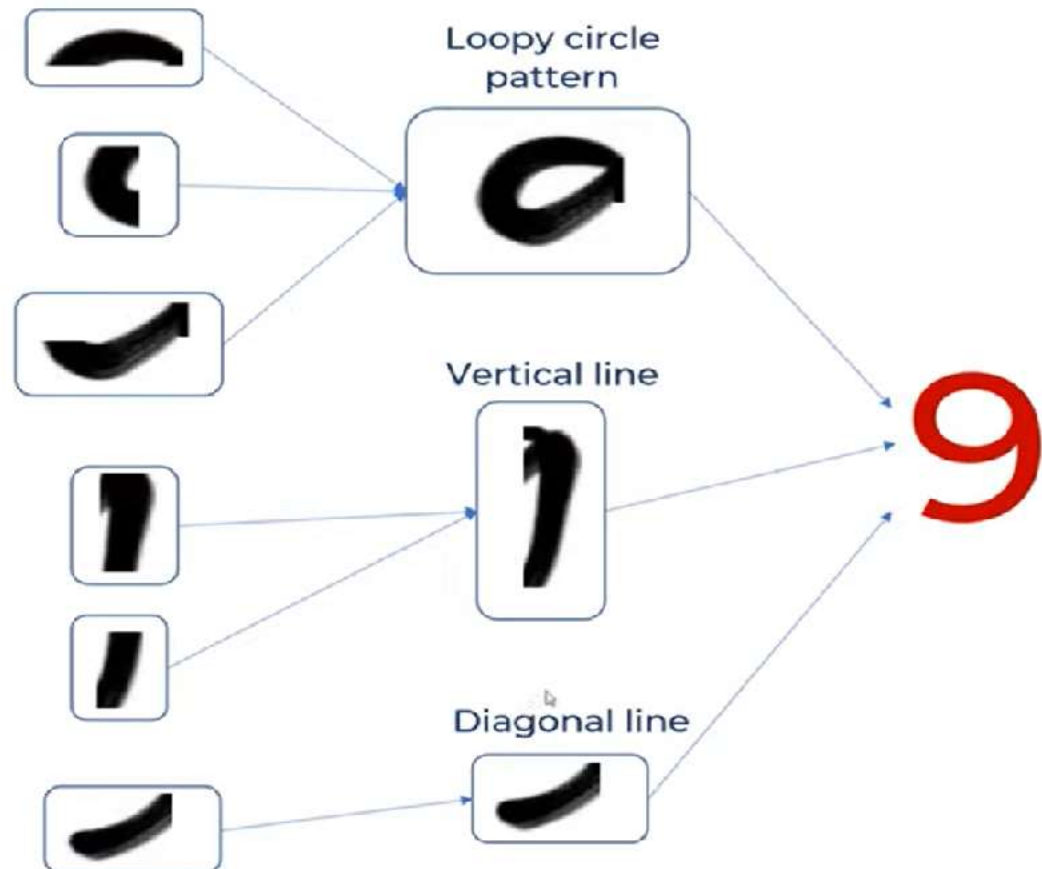


Koala's **body**? = Y

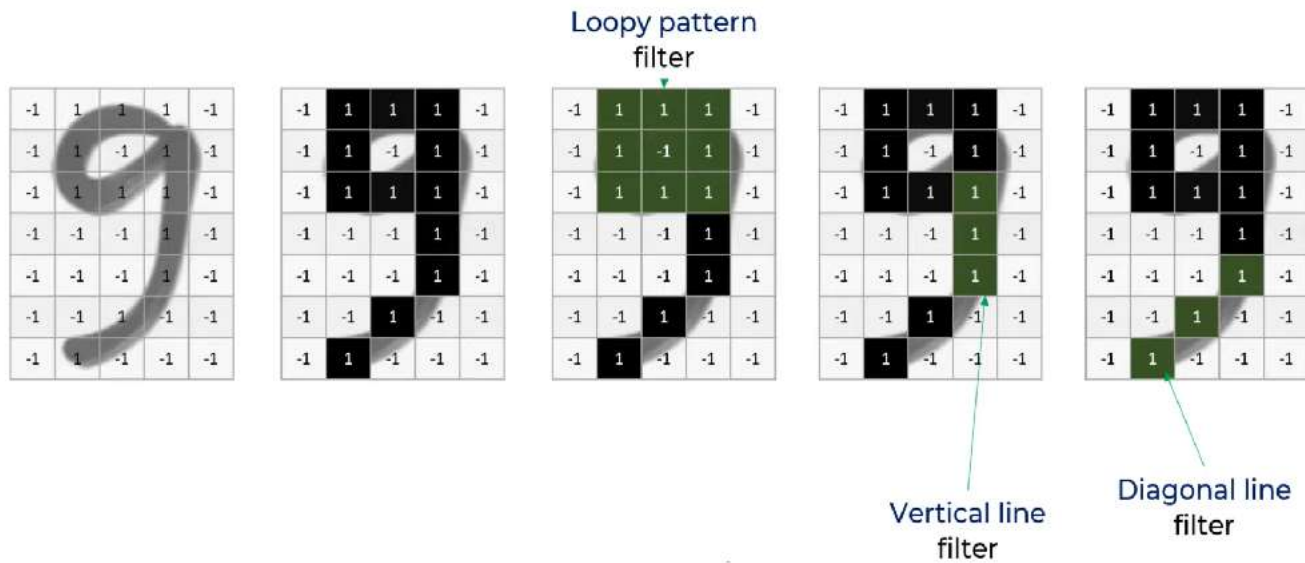


Is it **Koala**? = Y

Handwritten Digit



Handwritten Digit



Handwritten Digit

$$-1+1+1-1-1-1-1+1+1 = -1 \rightarrow -1/9 = -0.11$$

| | | | | |
|----|----------|----------|----------|----|
| -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | 1 | -1 |
| -1 | 1 | 1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | 1 | -1 | -1 |
| -1 | 1 | -1 | -1 | -1 |

*

| | | |
|----------|-----------|----------|
| 1 | 1 | 1 |
| 1 | -1 | 1 |
| 1 | 1 | 1 |

| | | |
|-------|--|--|
| -0.11 | | |
| | | |
| | | |
| | | |
| | | |

Handwritten Digit

| | | | | |
|----|----------|----------|----------|----|
| -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | 1 | -1 |
| -1 | 1 | 1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | 1 | -1 | -1 |
| -1 | 1 | -1 | -1 | -1 |

*

| | | |
|----------|-----------|----------|
| 1 | 1 | 1 |
| 1 | -1 | 1 |
| 1 | 1 | 1 |

| | | |
|-------|----------|--|
| -0.11 | 1 | |
| | | |
| | | |
| | | |
| | | |

Handwritten Digit

| | | | | |
|----|----|----|----|----|
| -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | 1 | -1 |
| -1 | 1 | 1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | -1 | 1 | -1 |
| -1 | -1 | 1 | -1 | -1 |
| -1 | 1 | -1 | -1 | -1 |

*

| | | |
|---|----|---|
| 1 | 1 | 1 |
| 1 | -1 | 1 |
| 1 | 1 | 1 |

| | | |
|-------|-------|-------|
| -0.11 | 1 | -0.11 |
| -0.55 | 0.11 | -0.33 |
| -0.33 | 0.33 | -0.33 |
| -0.22 | -0.11 | -0.22 |
| -0.33 | -0.33 | -0.33 |

Feature Map

Handwritten Digit

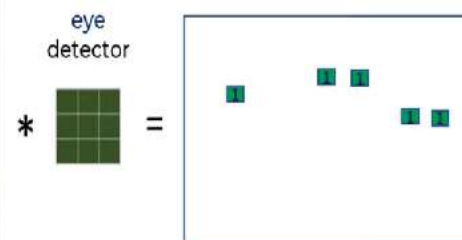
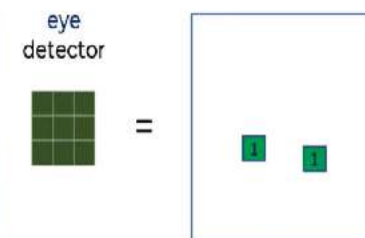
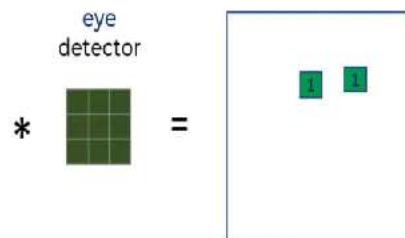
9 * $\begin{matrix} \text{Loopy pattern} \\ \text{detector} \\ \begin{matrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & 1 \end{matrix} \end{matrix} = \begin{matrix} 1 & & \\ & & \\ & & \\ & & \\ & & \end{matrix}$

6 * $\begin{matrix} \text{Loopy pattern} \\ \text{detector} \\ \begin{matrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & 1 \end{matrix} \end{matrix} = \begin{matrix} & & & \\ & & & \\ & & & \\ & & & \\ & & 1 & \end{matrix}$

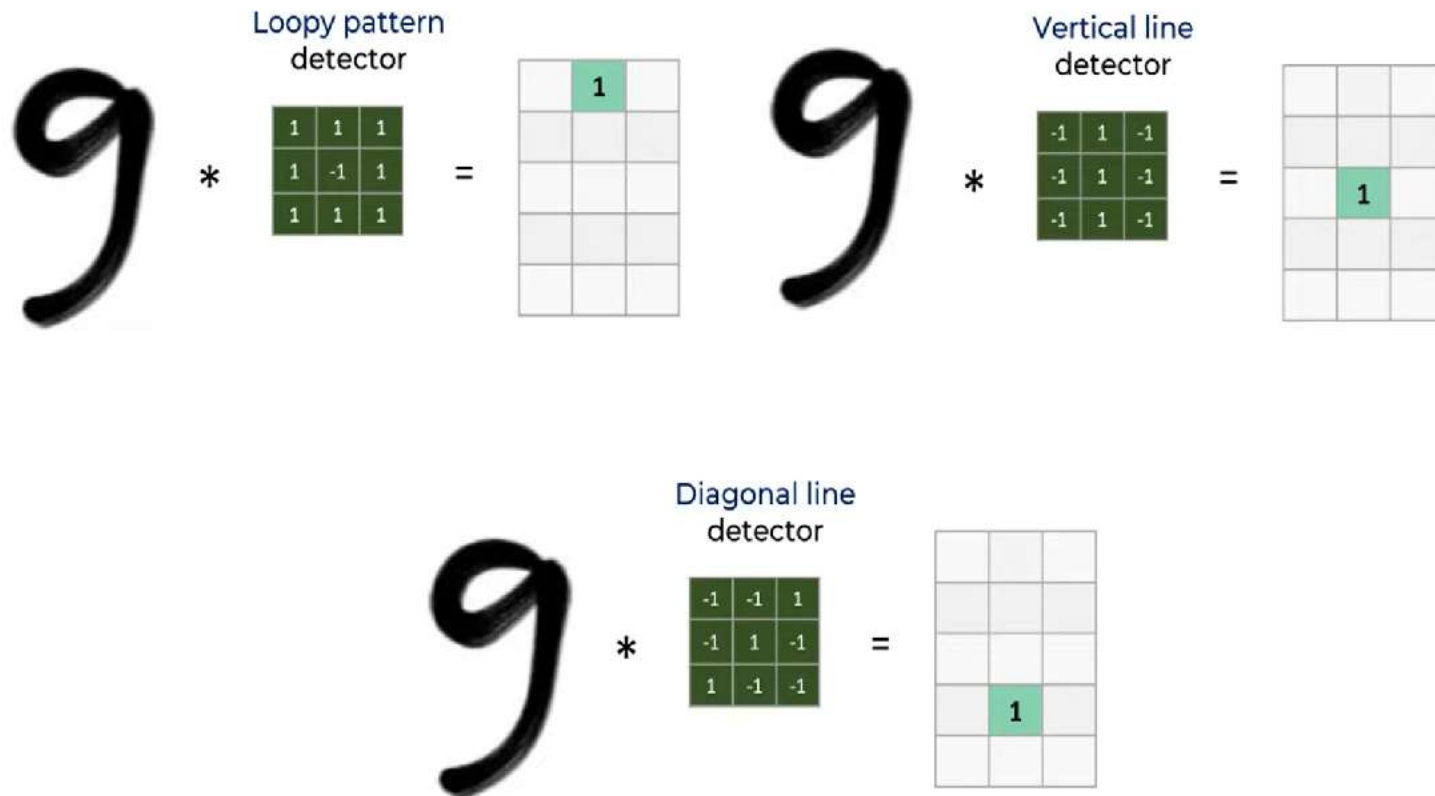
8 * $\begin{matrix} \text{Loopy pattern} \\ \text{detector} \\ \begin{matrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & 1 \end{matrix} \end{matrix} = \begin{matrix} & & 1 & \\ & & & \\ & & & \\ & & & \\ & 1 & & \end{matrix}$

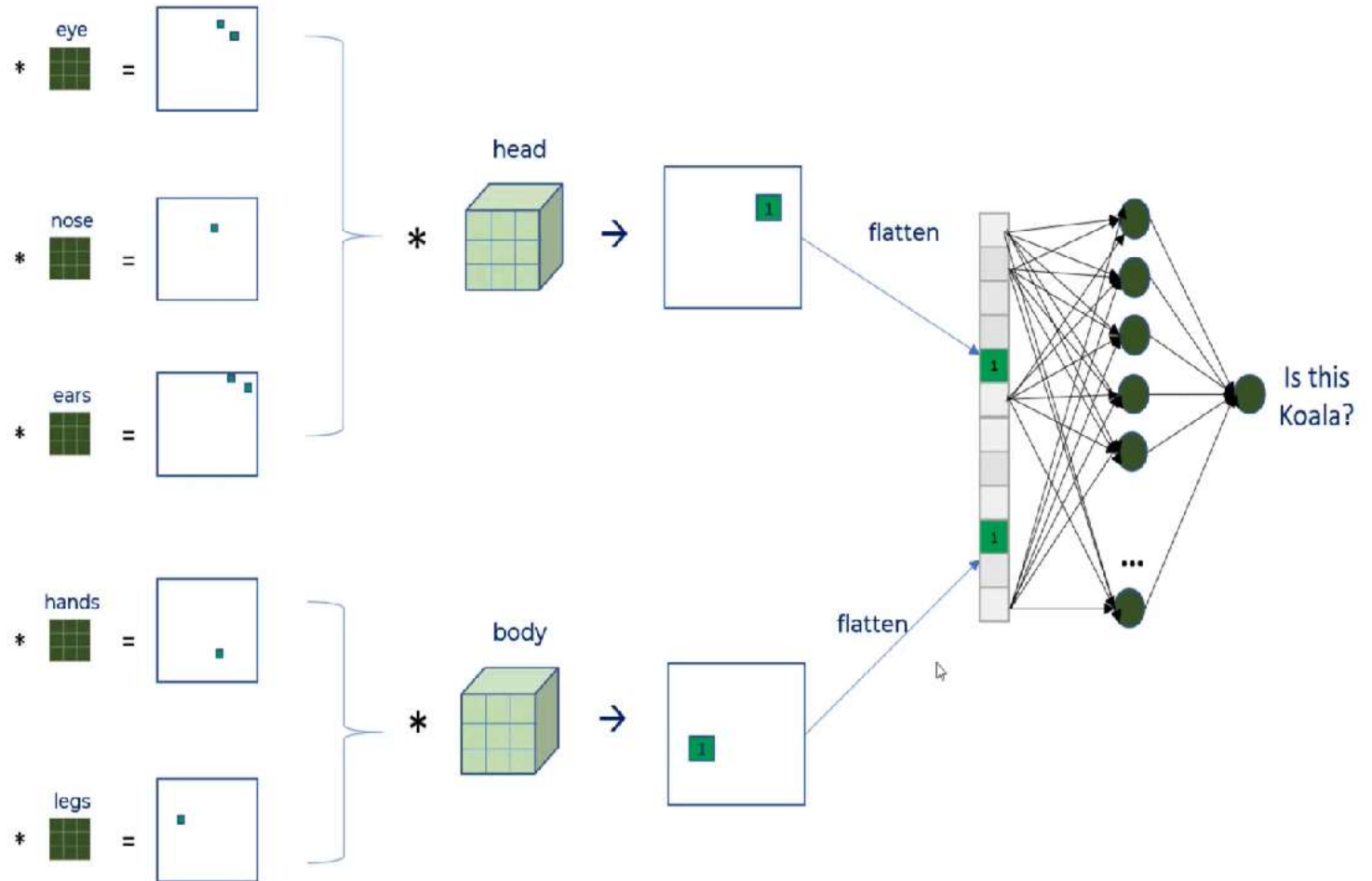
96 * $\begin{matrix} \text{Loopy pattern} \\ \text{detector} \\ \begin{matrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & 1 \end{matrix} \end{matrix} = \begin{matrix} 1 & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & 1 & \end{matrix}$

Handwritten Digit

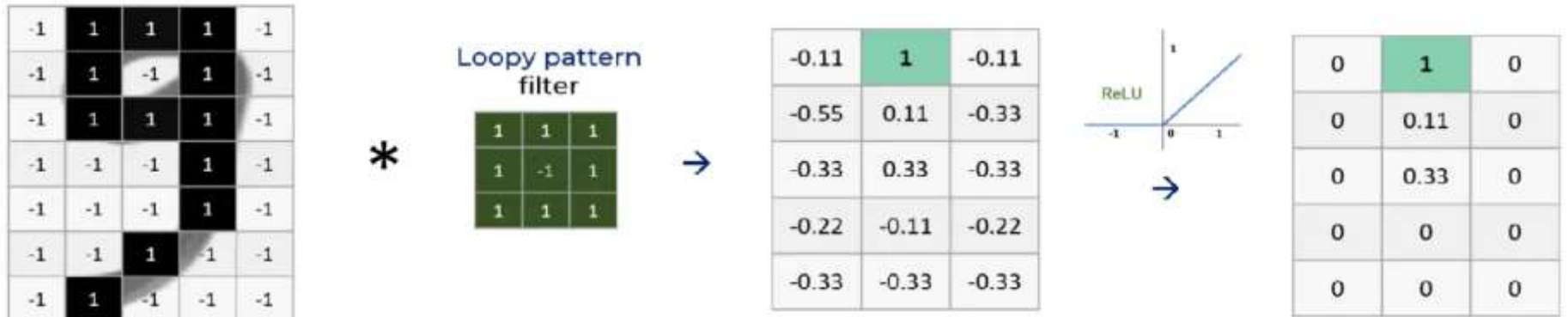


Handwritten Digit

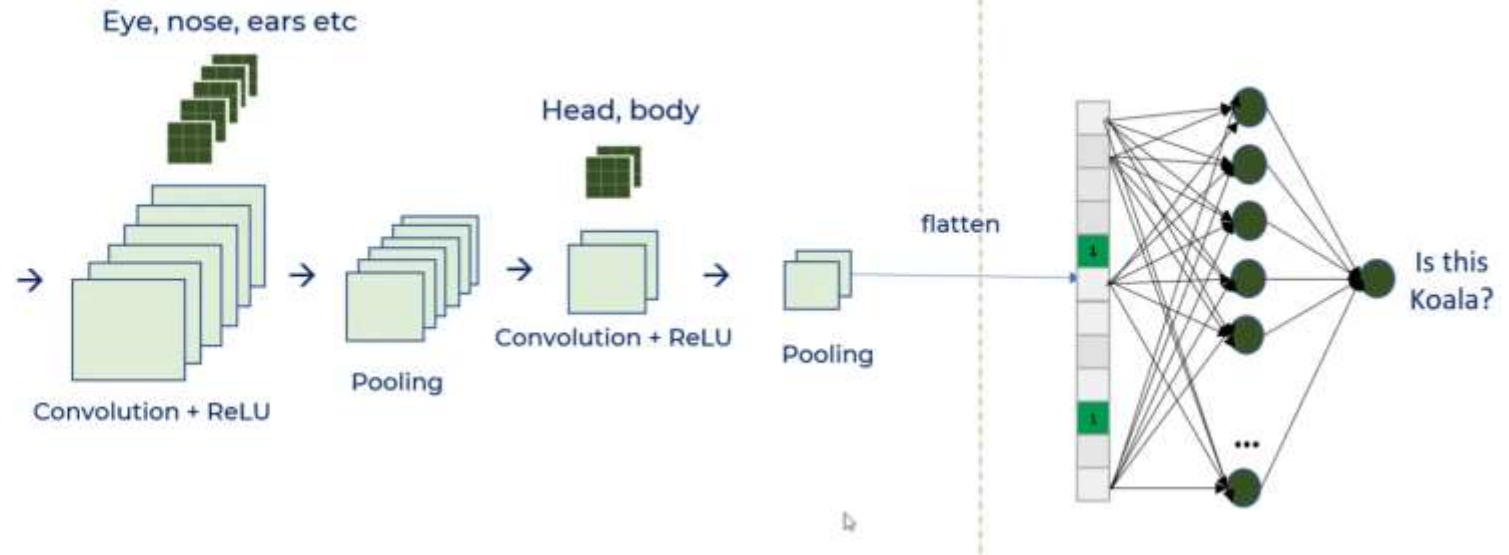




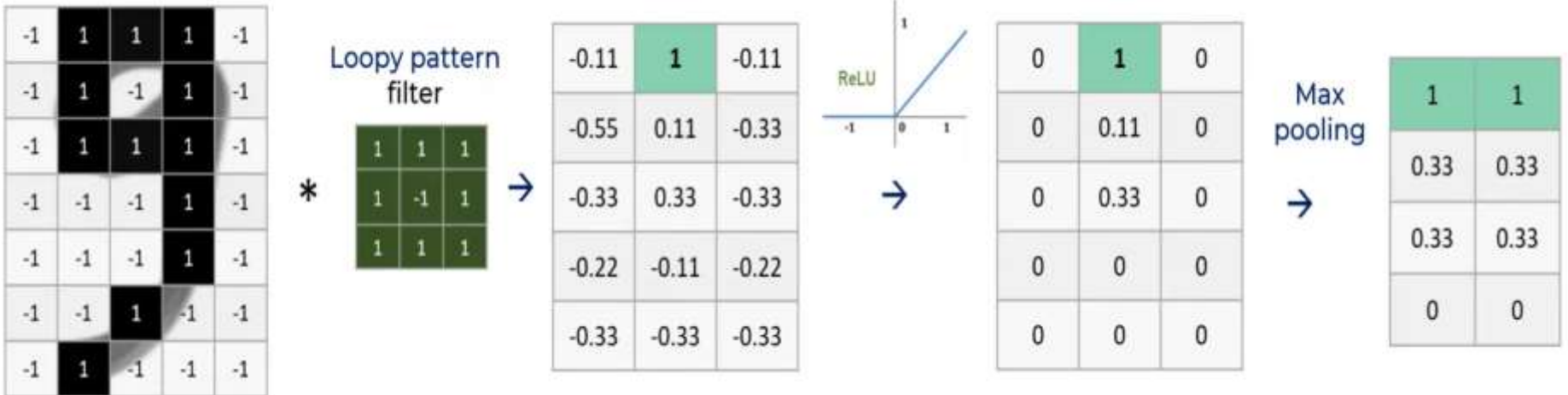
Handwritten Digit



Handwritten Digit



Handwritten Digit



Handwritten Digit

Shifted 9 at
different position

| | | | | |
|----|----|----|----|----|
| 1 | 1 | 1 | -1 | -1 |
| 1 | -1 | 1 | -1 | -1 |
| 1 | 1 | 1 | -1 | -1 |
| -1 | -1 | 1 | -1 | -1 |
| -1 | -1 | 1 | -1 | -1 |
| -1 | 1 | -1 | -1 | -1 |
| 1 | -1 | -1 | -1 | -1 |

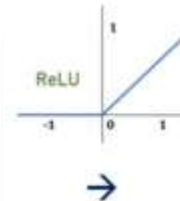
Loopy pattern
filter

*

| | | |
|---|----|---|
| 1 | 1 | 1 |
| 1 | -1 | 1 |
| 1 | 1 | 1 |

→

| | | |
|-------|-------|-------|
| 1 | -0.11 | -0.11 |
| 0.11 | -0.33 | 0.33 |
| 0.33 | -0.33 | -0.33 |
| -0.11 | -0.55 | -0.33 |
| -0.55 | -0.33 | -0.55 |



→

| | | |
|------|---|------|
| 1 | 0 | 0 |
| 0.11 | 0 | 0.33 |
| 0.33 | 0 | 0 |
| 0 | 0 | 0 |
| 0 | 0 | 0 |

Max
pooling

→

| | |
|------|------|
| 1 | 0.33 |
| 0.33 | 0.33 |
| 0.33 | 0 |
| 0 | 0 |



THANK YOU