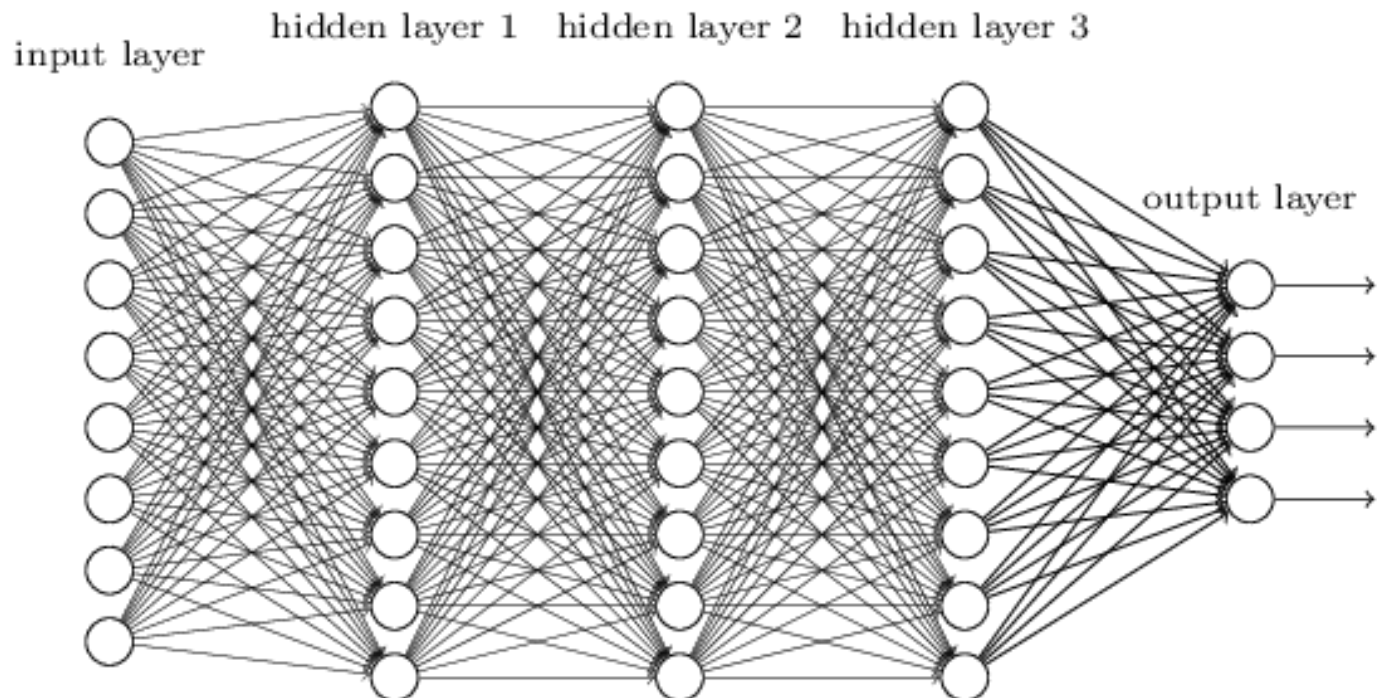


- We know it is good to learn a small model.
- From this fully connected model, do we really need all the edges?
- Can some of these be shared?



Consider learning an image:

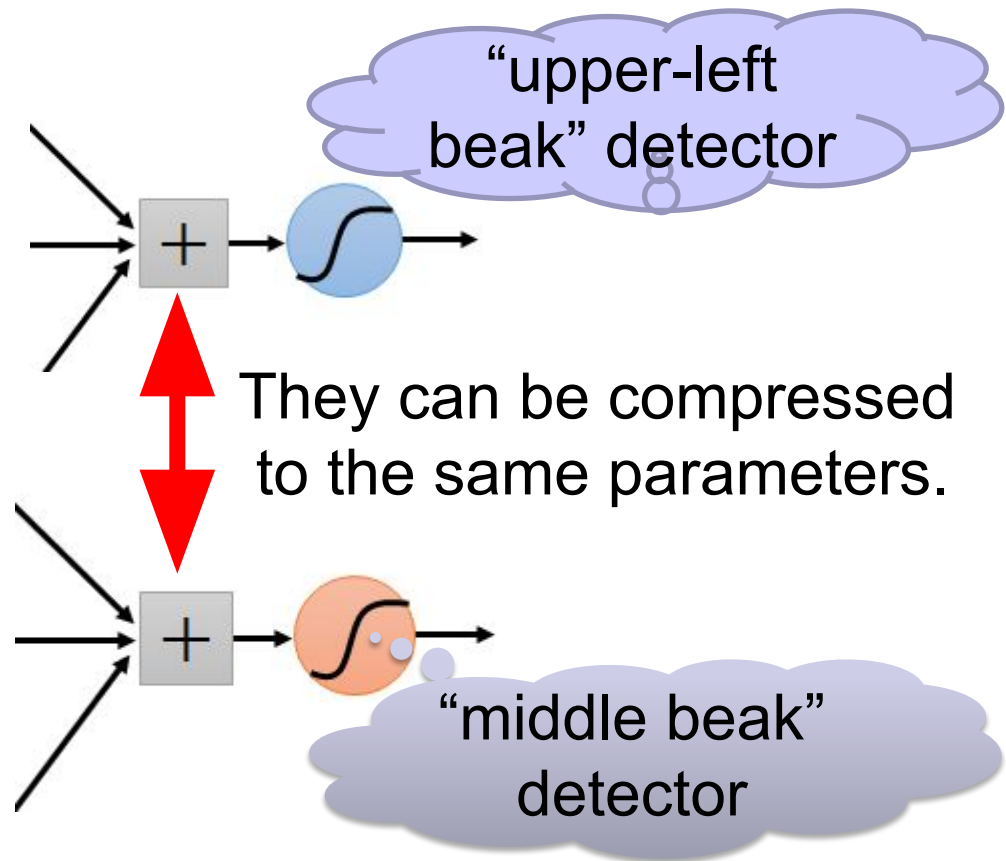
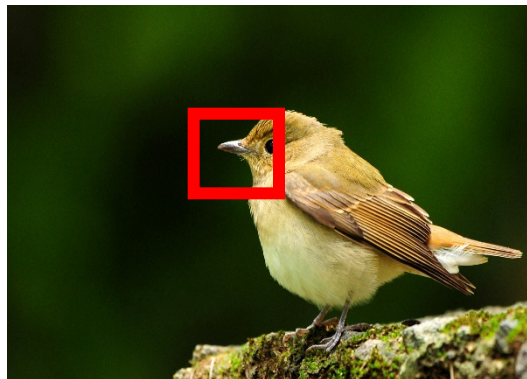
- Some patterns are much smaller than the whole image

Can represent a small region with fewer parameters



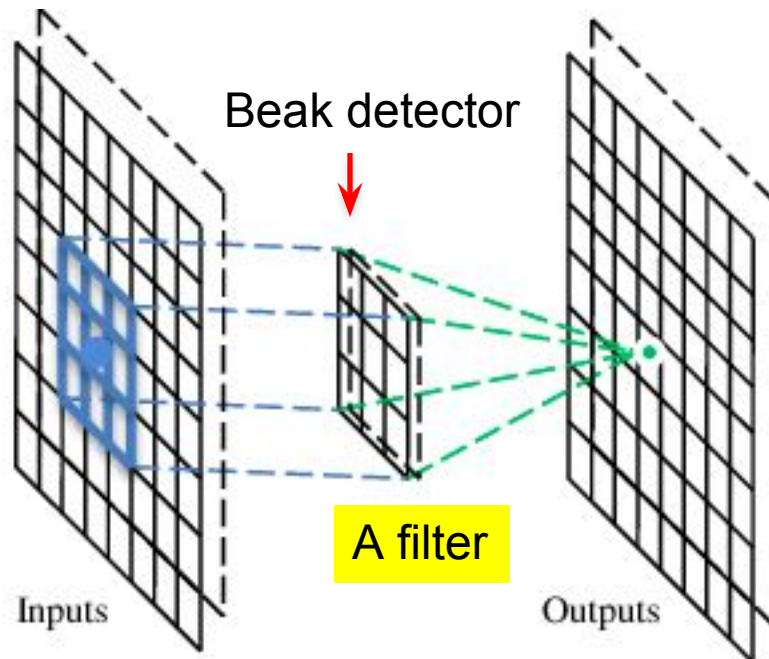
Same pattern appears in different places:
They can be compressed!

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

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

These are the network parameters to be learned.

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

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

Dot
product



3

-1

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

Filter 1

Convolution

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

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

Filter 1

3

-3

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 1

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

Convolution

stride=

1

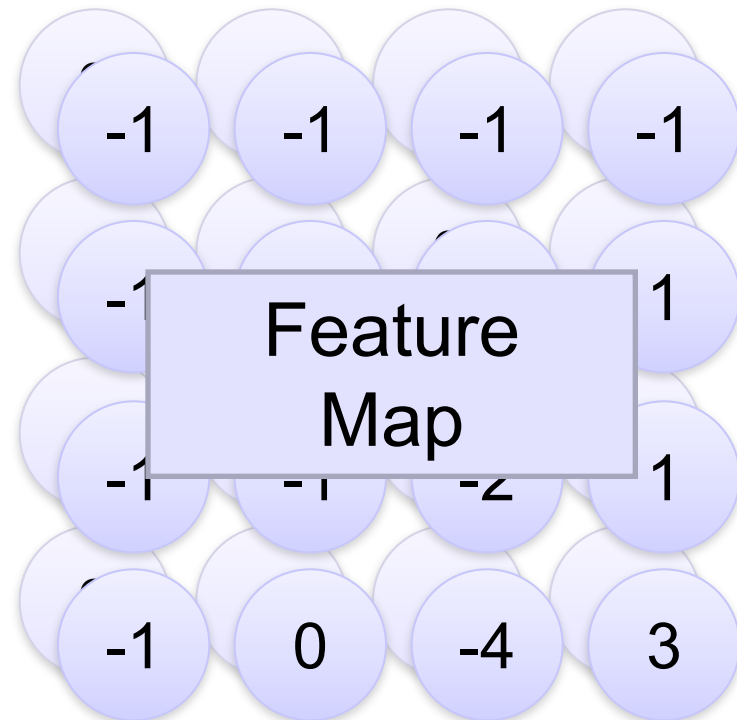
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

6 x 6 image

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

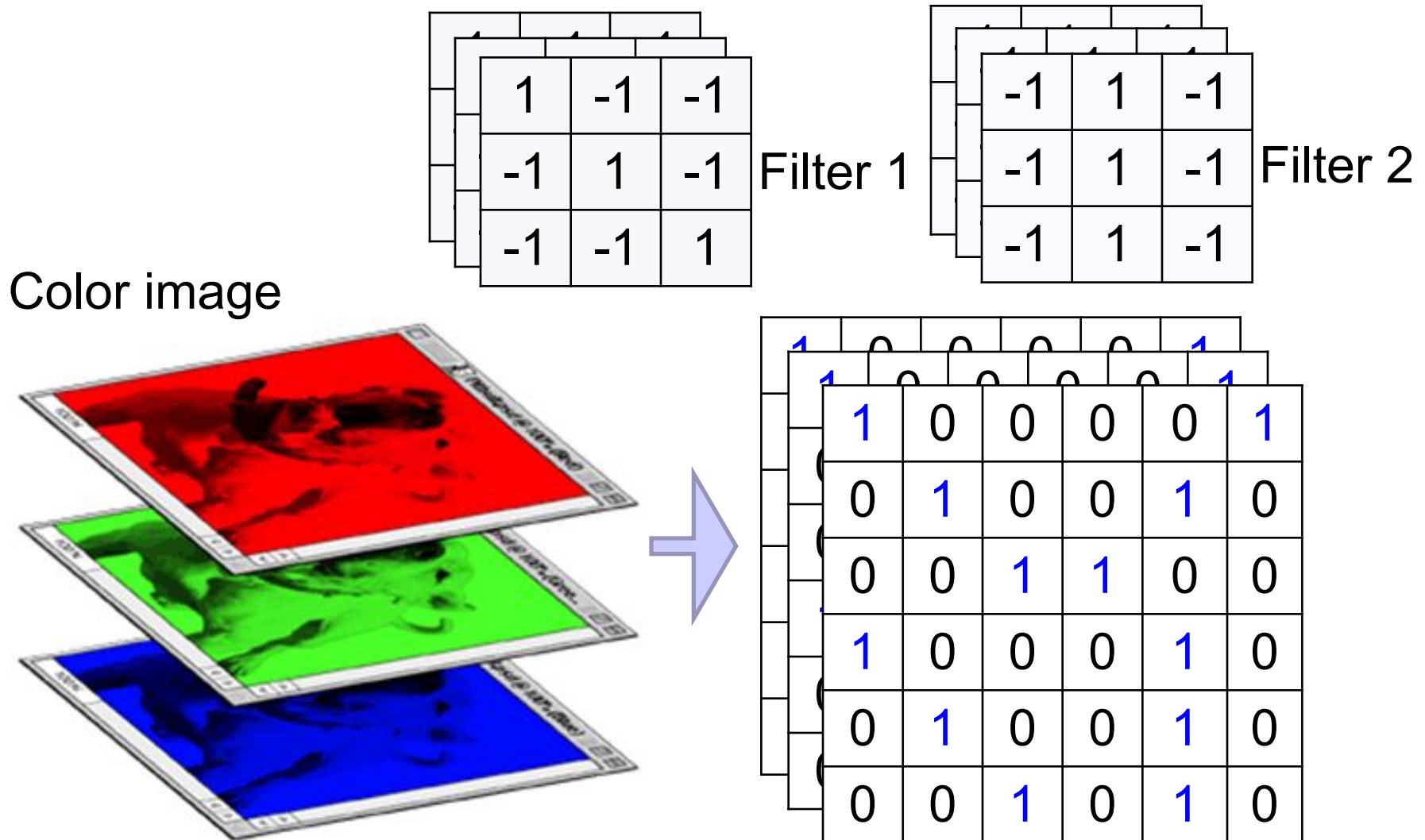
Filter 2

Repeat this for each filter



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

Color image: RGB 3 channels



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

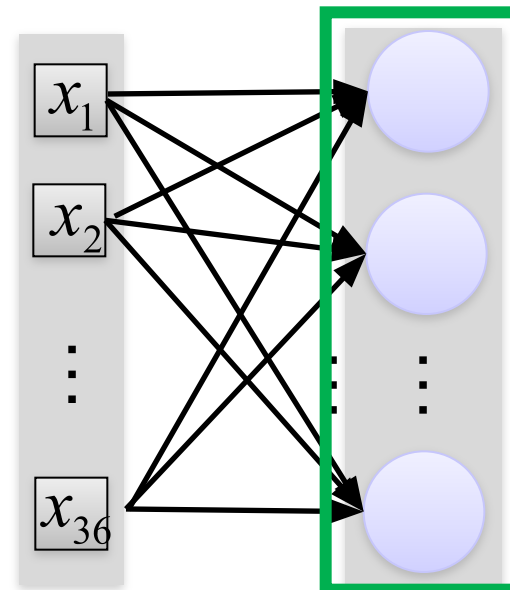


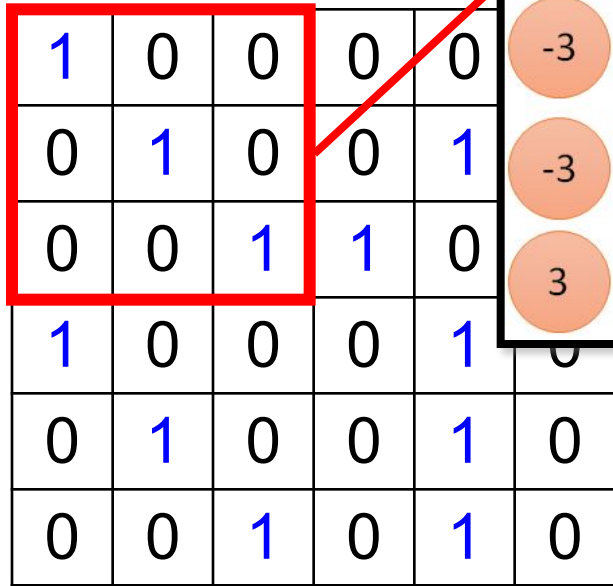
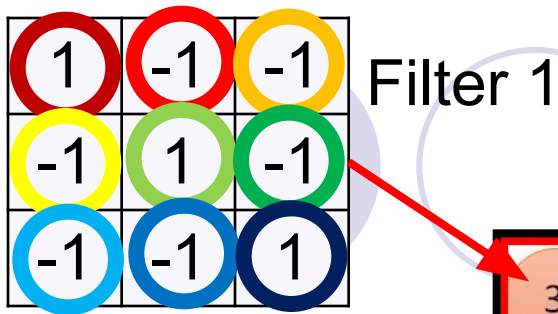
convolution

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

Fully-connected

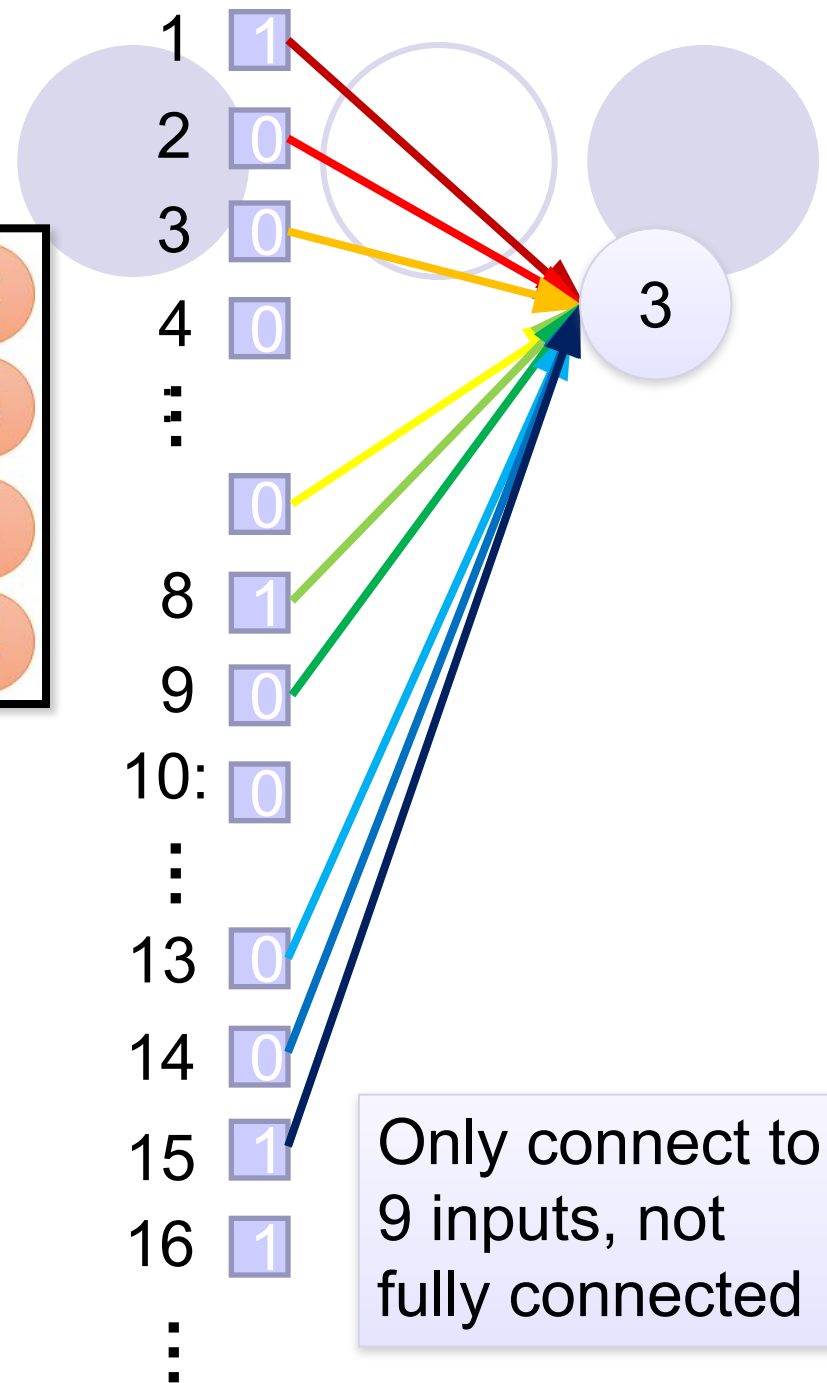
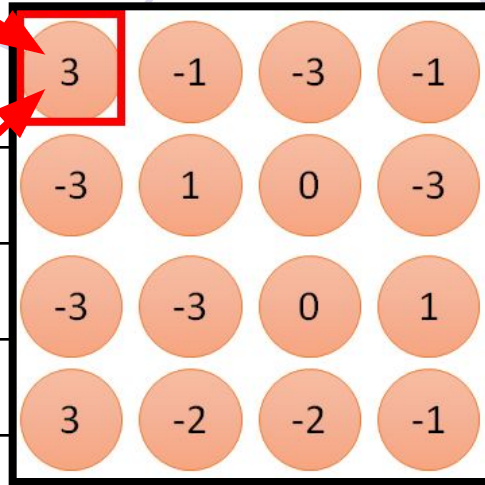
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

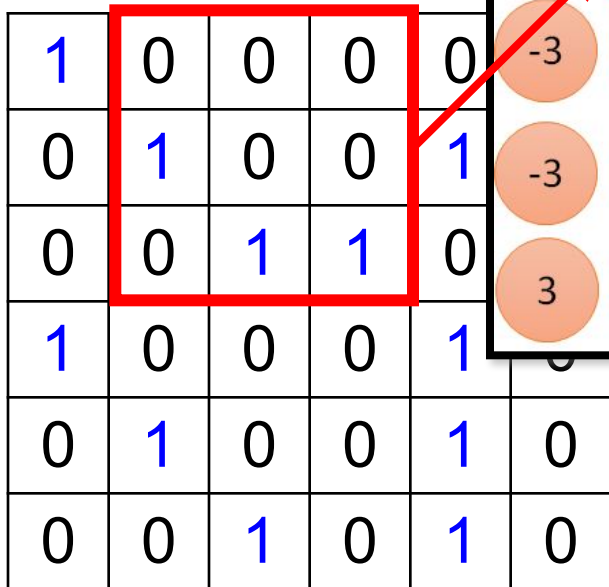
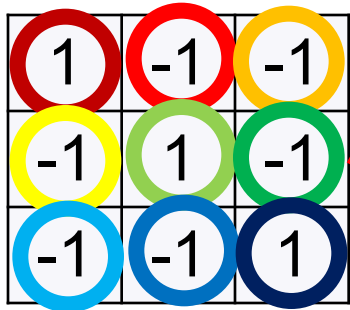




6 x 6 image

fewer parameters!

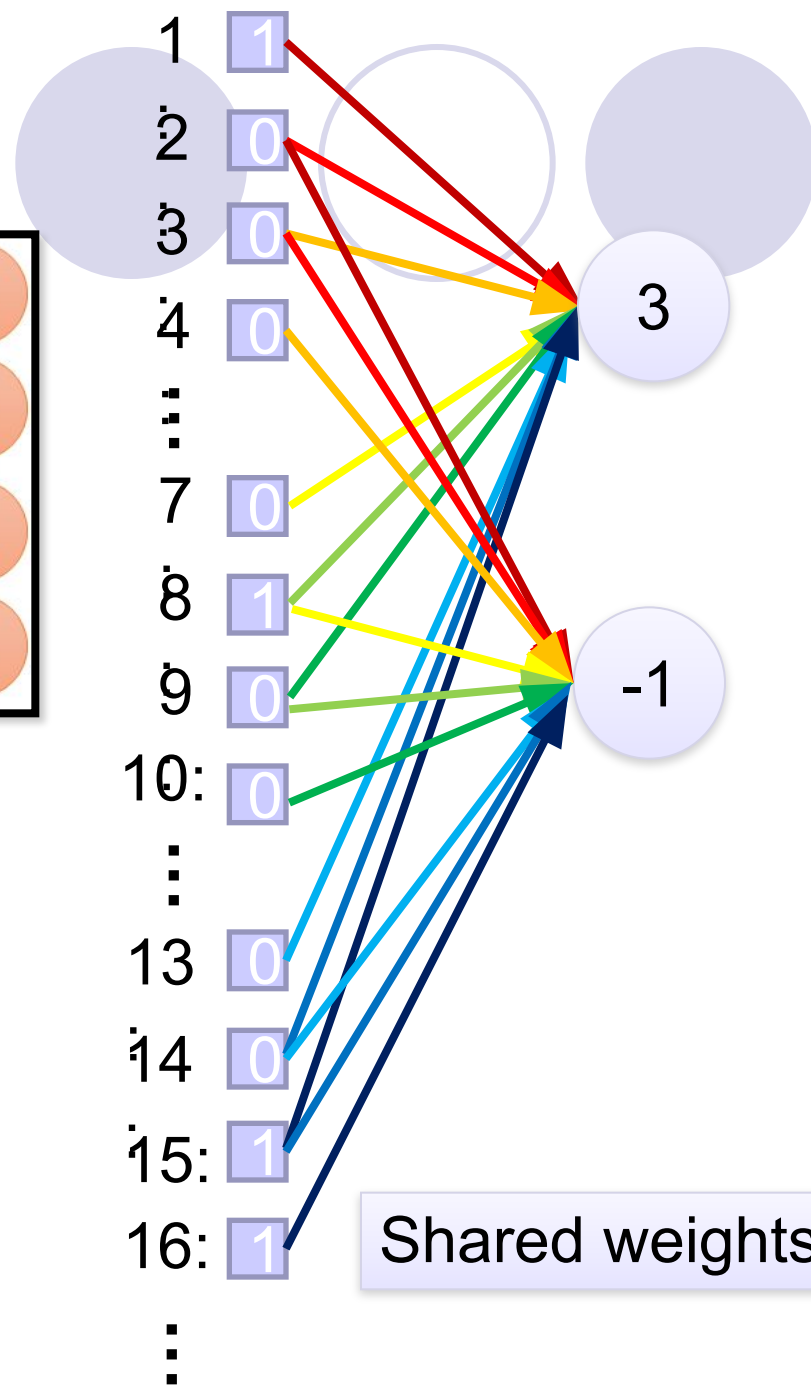
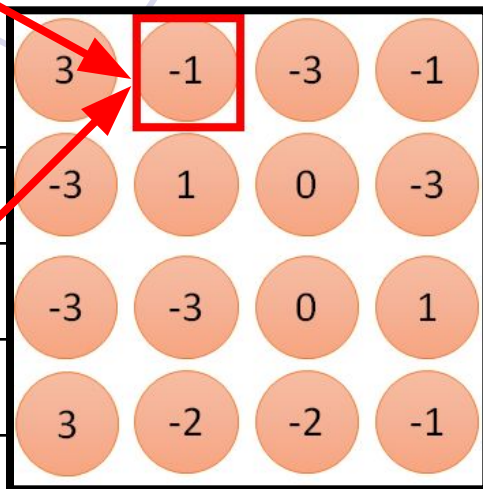




6 x 6 image

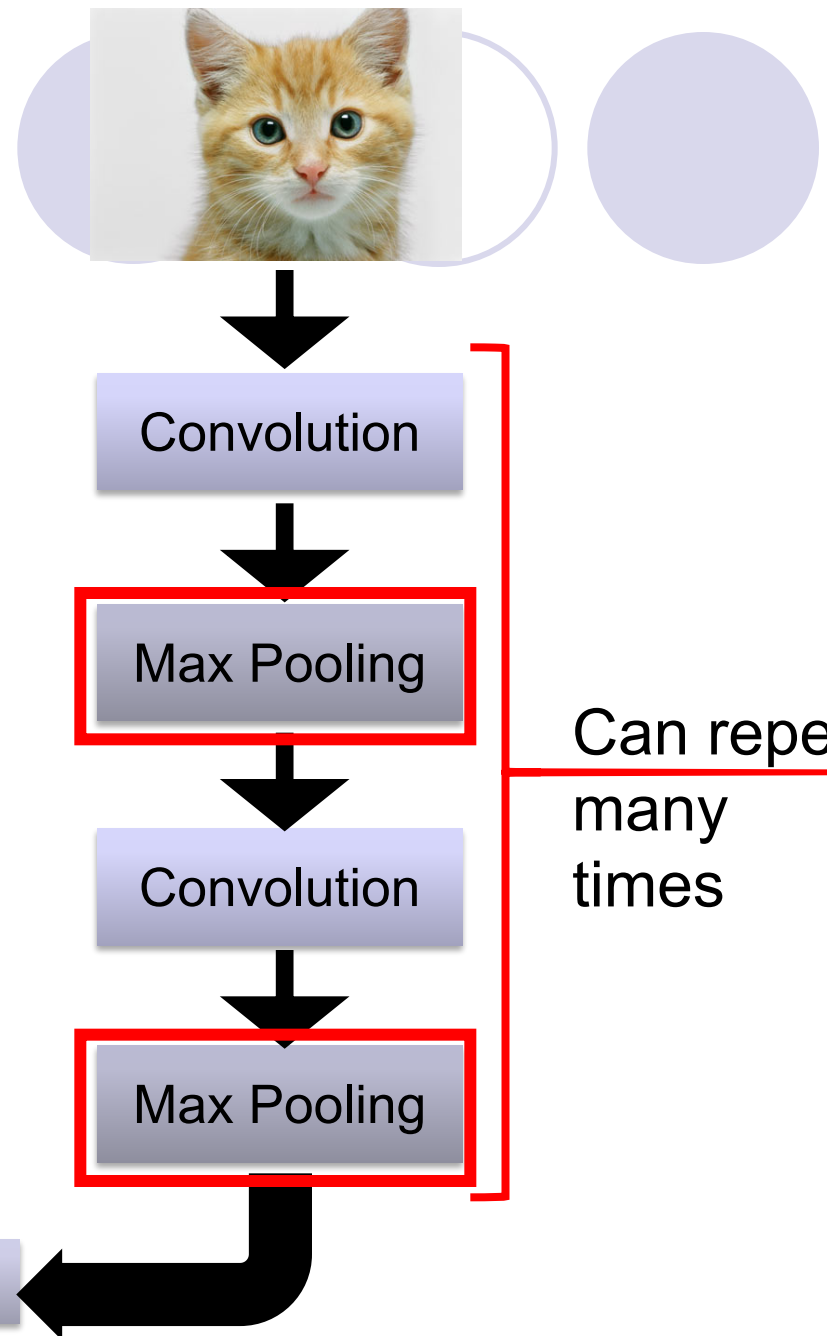
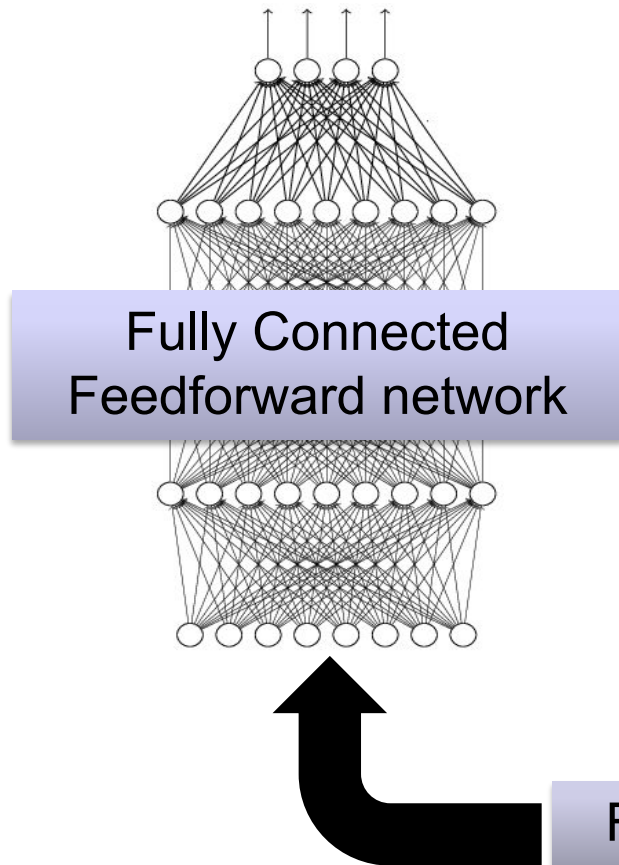
Fewer parameters

Even fewer parameters



The whole CNN

cat dog



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 smaller

→ fewer parameters to characterize the image



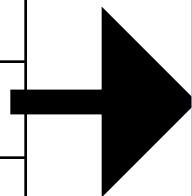
A CNN compresses a fully connected network in two ways:

- 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

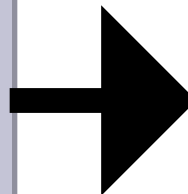
6 x 6 image



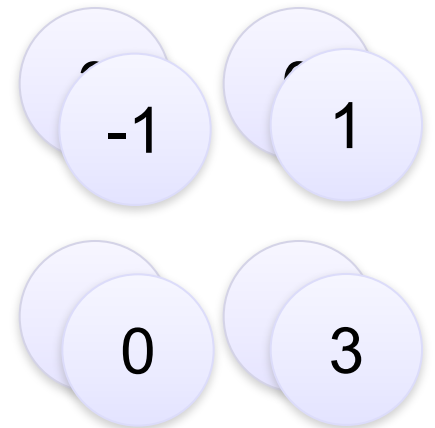
Conv



Max
Pooling



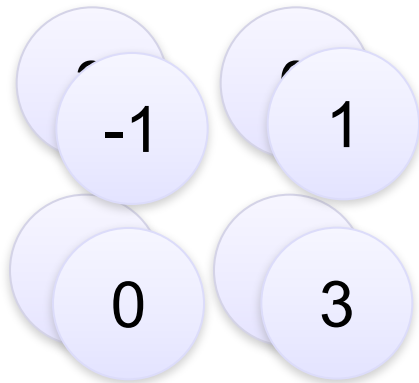
New image
but smaller



2 x 2 image

Each filter
is a channel

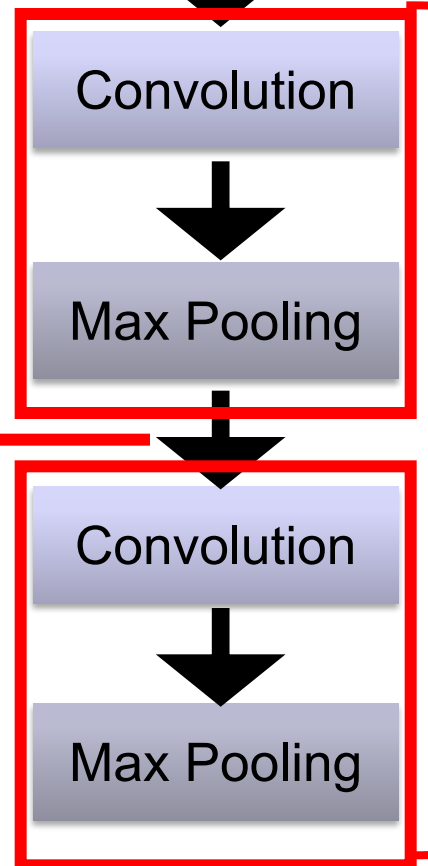
The whole CNN



A new image

Smaller than the original image

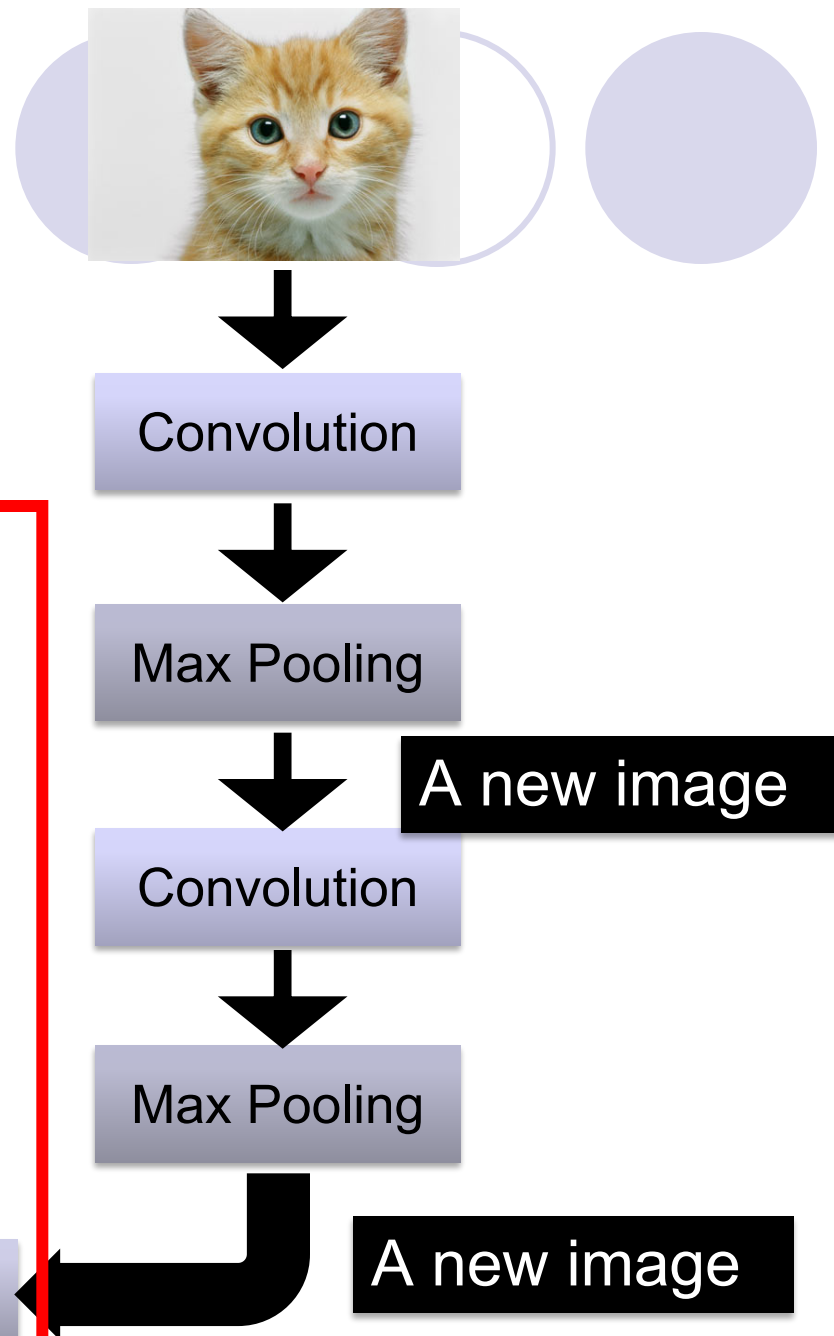
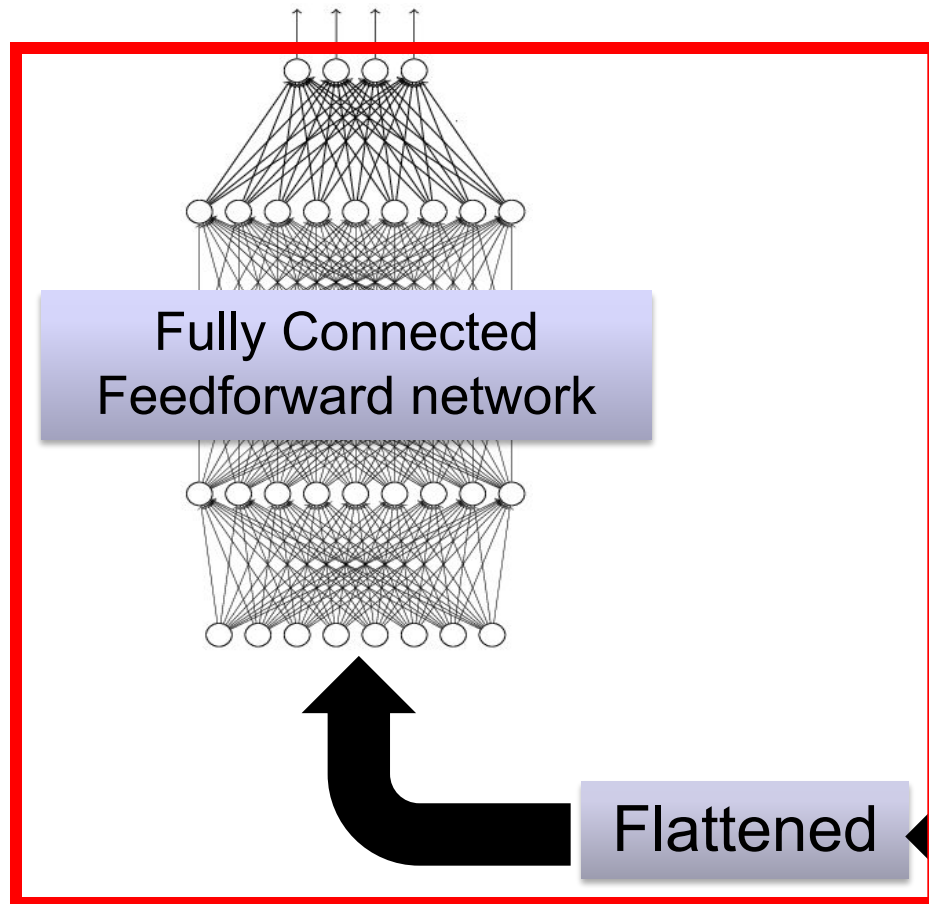
The number of channels is the number of filters



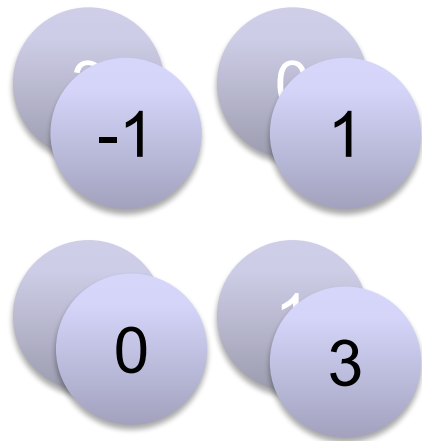
Can repeat many times

The whole CNN

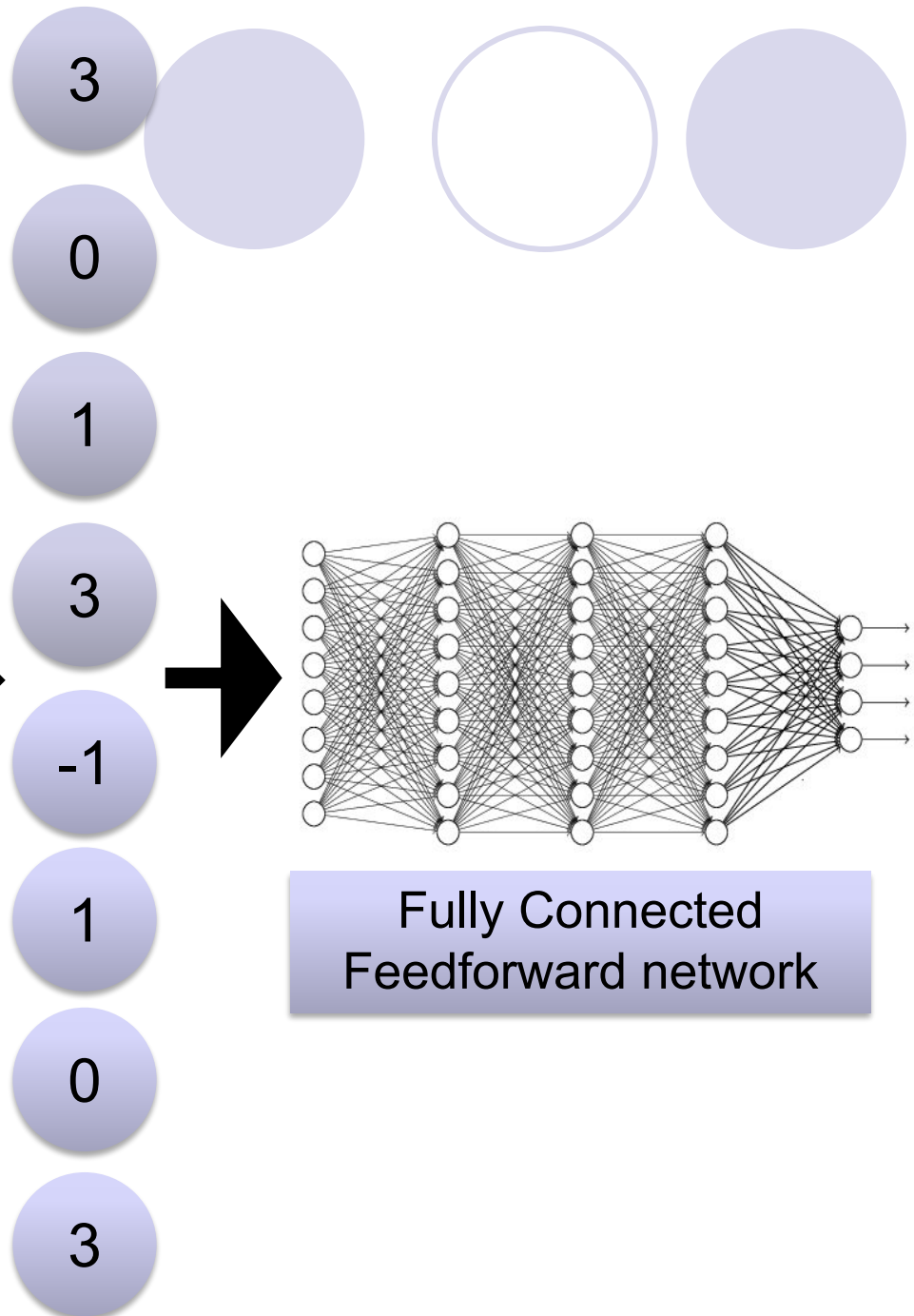
cat dog



Flattening



Flattened



Fully Connected
Feedforward network

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
-1	-1	-1	1	-1

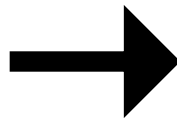
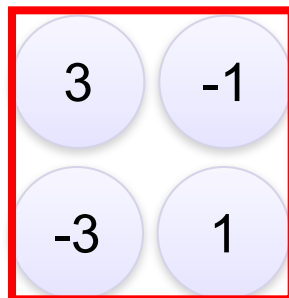
There are
25 3x3
filters.

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)**

Input



Convolution



Max Pooling



Convolution



Max Pooling

1 x 28 x 28

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

How many parameters for each filter?

9

25 x 26 x 26

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

25 x 13 x 13

```
model2.add(Convolution2D(50, 3, 3))
```

How many parameters for each filter?

225=
25x9

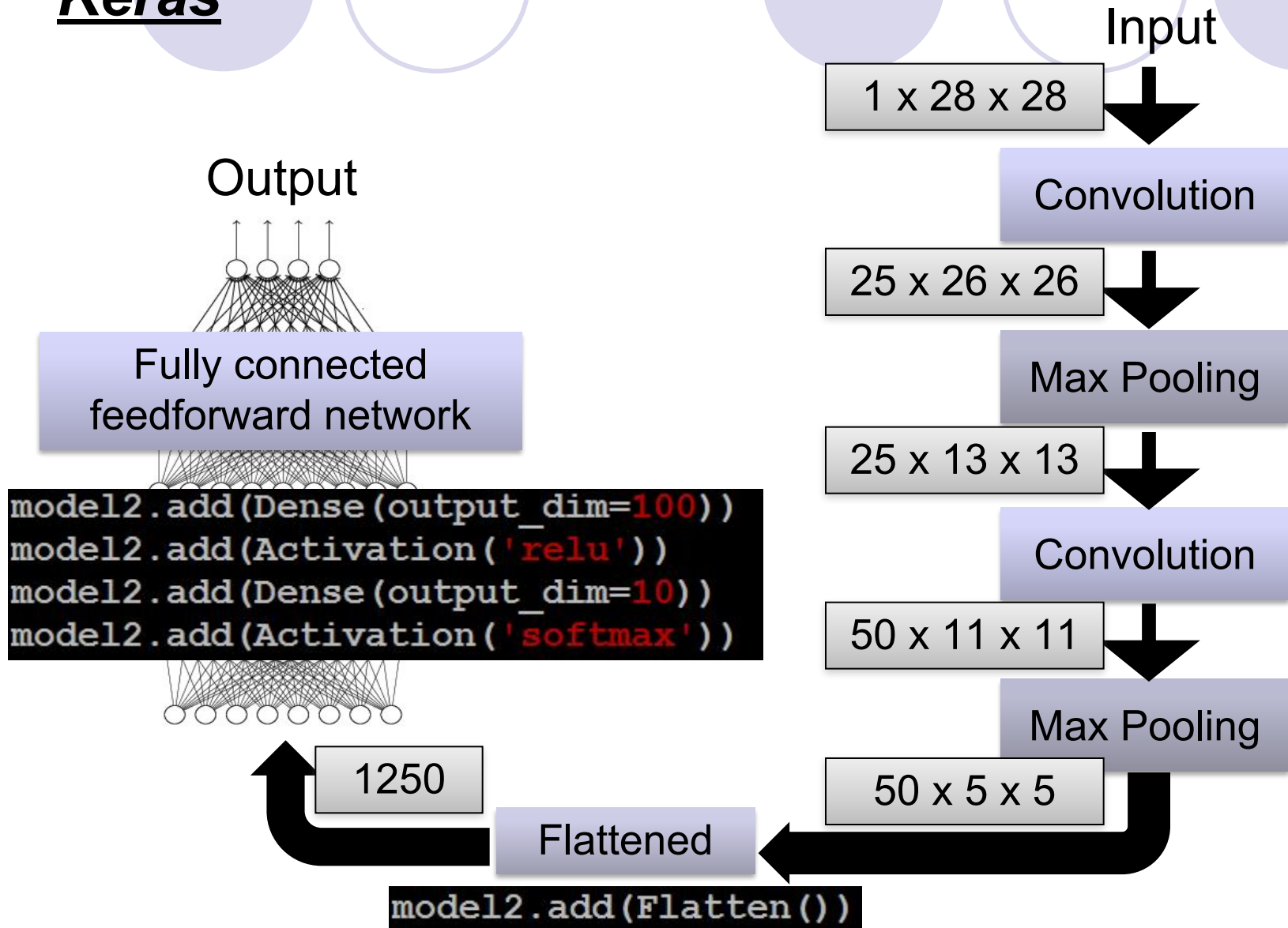
50 x 11 x 11

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

50 x 5 x 5

CNN in Keras

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



AlphaGo



19 x 19 matrix

Black: 1
white: -1
none: 0

Neural
Network

Next move
(19 x 19
positions)

Fully-connected feedforward
network can be used

But CNN performs much better

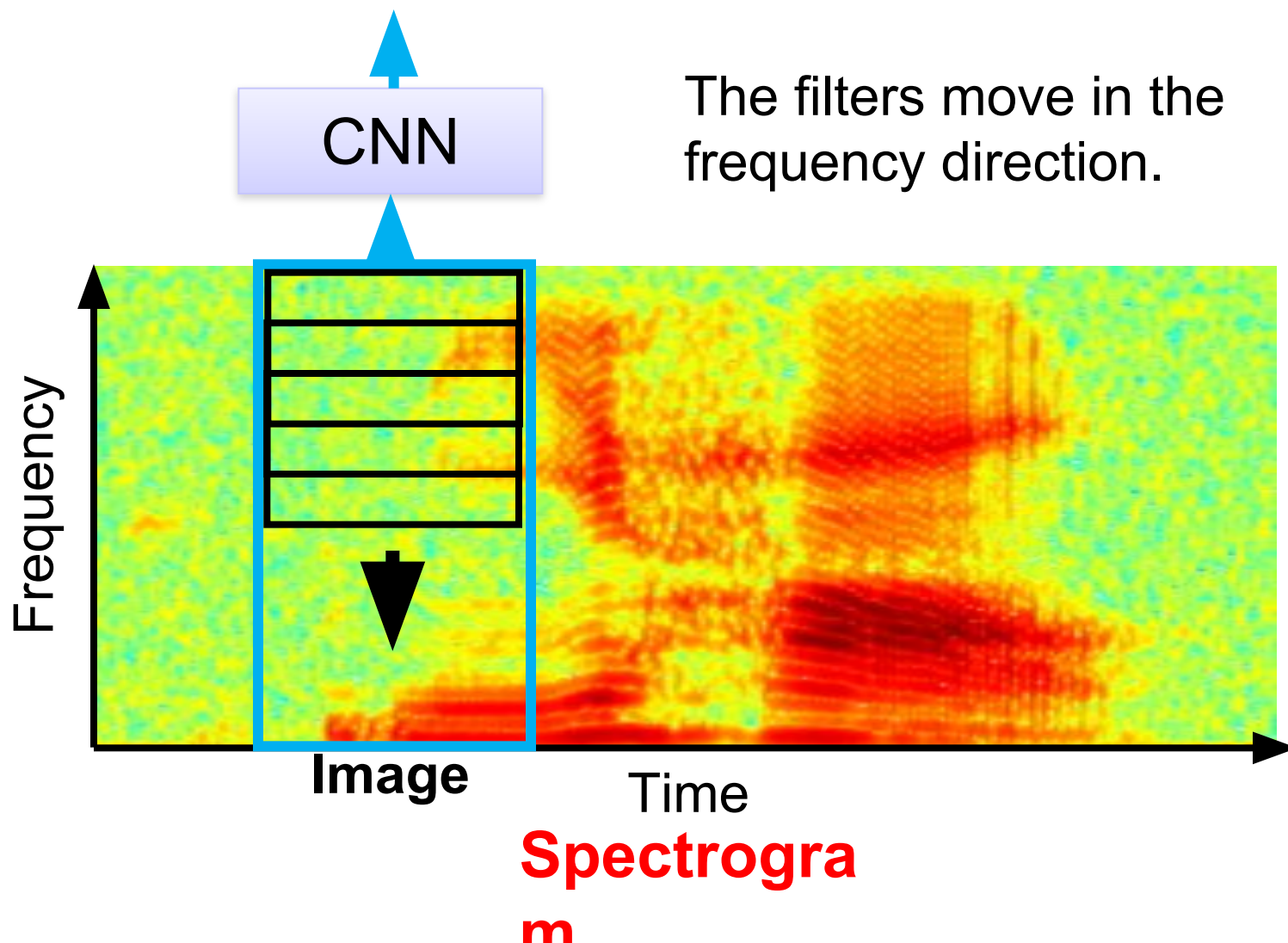
AlphaGo's policy network

The following is quotation from their Nature article:

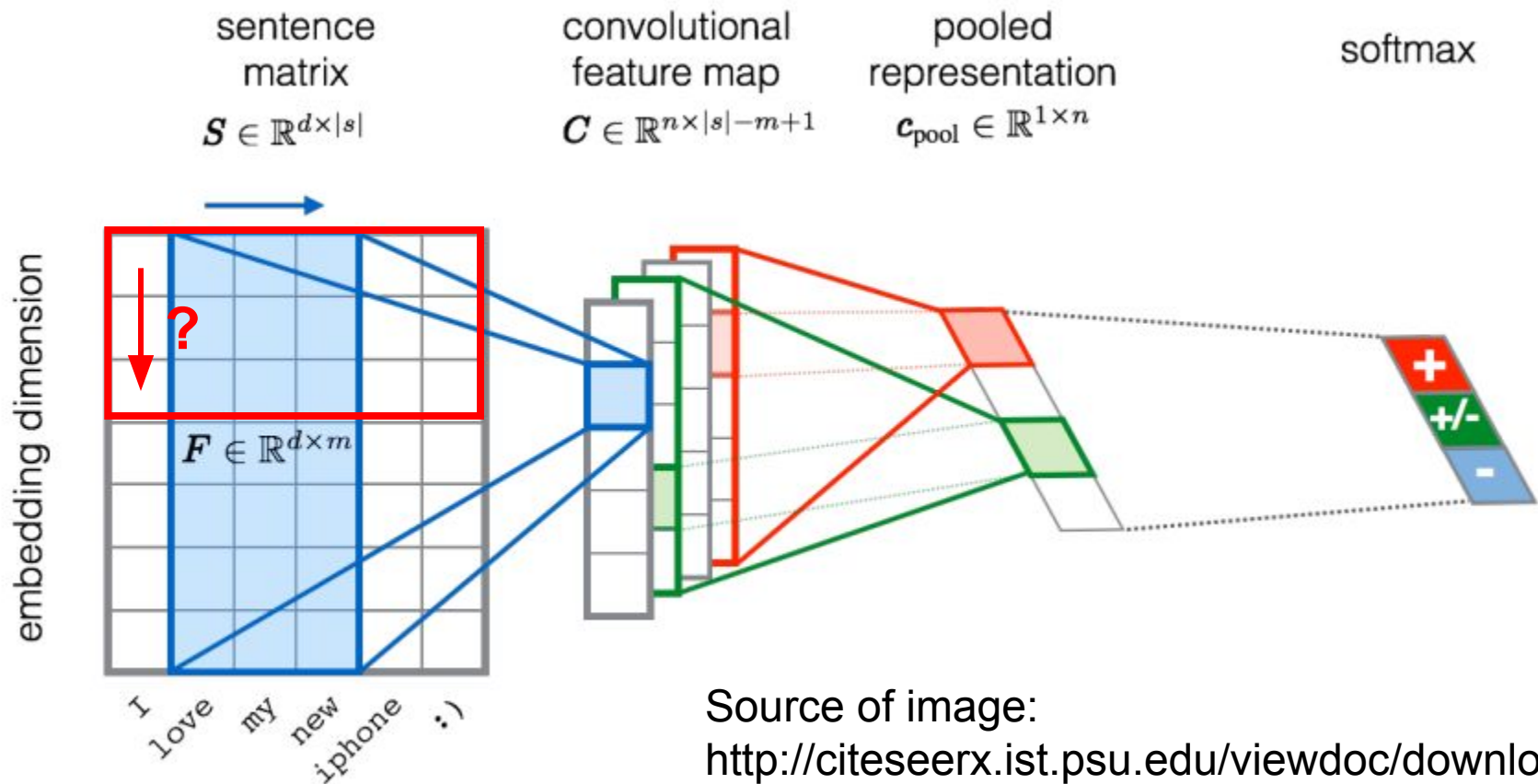
Note: AlphaGo does not use Max Pooling.

Neural network architecture. The input to the policy network is a $19 \times 19 \times 48$ image stack consisting of 48 feature planes. The first hidden layer zero pads the input into a 23×23 image, then convolves k filters of kernel size 5×5 with stride 1 with the input image and applies a rectifier nonlinearity. Each of the subsequent hidden layers 2 to 12 zero pads the respective previous hidden layer into a 21×21 image, then convolves k filters of kernel size 3×3 with stride 1, again followed by a rectifier nonlinearity. The final layer convolves 1 filter of kernel size 1×1 with stride 1, with a different bias for each position, and applies a softmax function. The match version of AlphaGo used $k = 192$ filters; Fig. 2b and Extended Data Table 3 additionally show the results of training with $k = 128, 256$ and 384 filters.

CNN in speech recognition



CNN in text classification



Source of image:

<http://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.703.6858&rep=rep1&type=pdf>