

# Convolutional Neural Networks

## CS-477 Computer Vision

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## 1 Introduction

## 2 Convolutional Neural Network

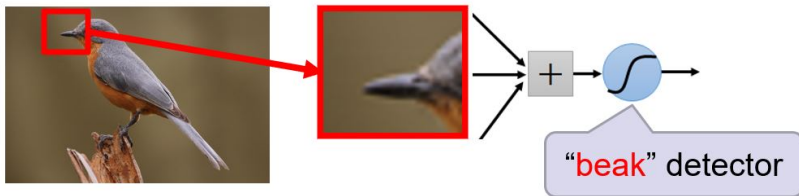
## 2 Convolutional Neural Network

- A convolutional neural network (CNN) is a type of artificial neural network used primarily for image recognition and processing, due to its ability to recognize patterns in images.
- FNN could not scale up to image and video processing tasks.
- CNN specifically tailored for image and video processing tasks.

Consider learning an image:

Some patterns are much smaller than the whole image

Can represent a small region with fewer parameters

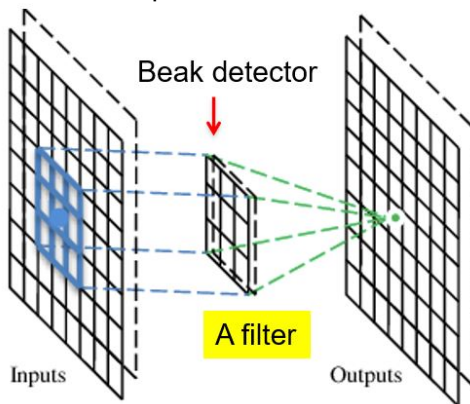


- They can be compressed!
- What about training a lot of such "small" detectors and each detector must "move around".



## 2

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.

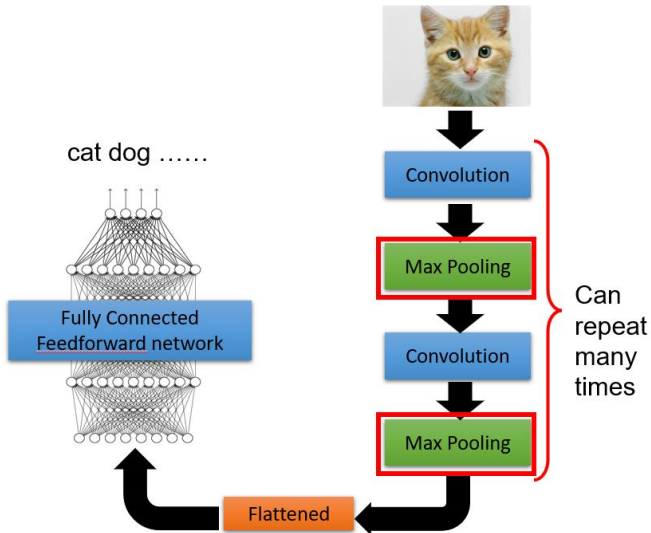




- Input layer
- Convolution layer
- Pooling layer
- Fully connected layer

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- Convolution layer
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- Fully connected layer

# Architecture of CNN



# Input layer

- Example input a 28 pixel by 28 pixel grayscale image
- Unlike FNN, we do not "flatten" the input to a 1D vector
  - Input is presented to network in 2D as 28 x 28 matrix

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

6 x 6 image

## Filter 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

Dot  
product  
→

3

-1

Filter 1

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

6 x 6 image

## 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



Filter 1

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

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

## Convolution

stride=1

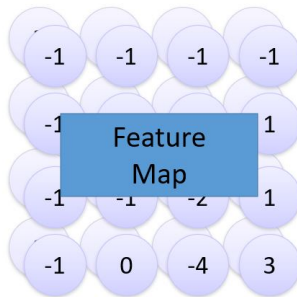
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0

6 x 6 image

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

Filter 2

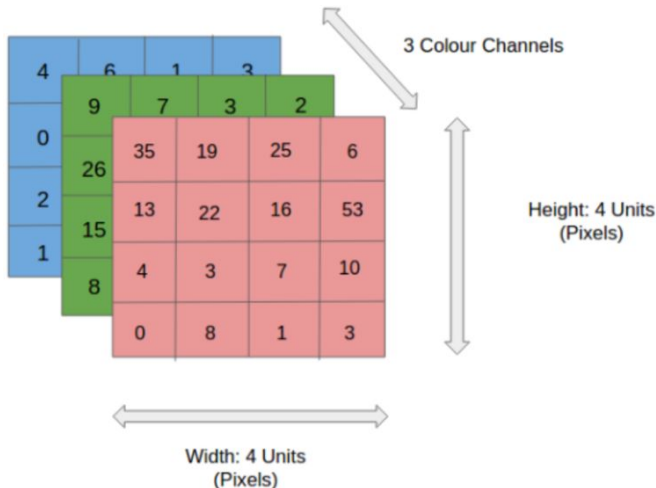
Repeat this for each filter



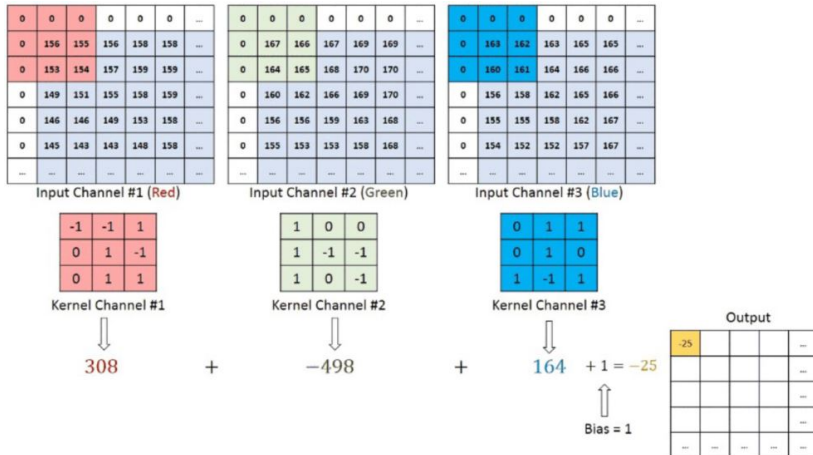


# Convolution in RGB

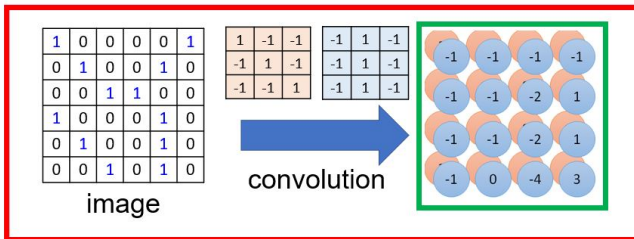
An RGB image is of the form



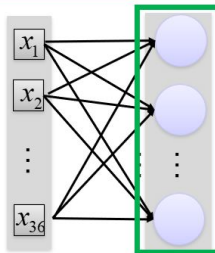
## Convolution in RGB



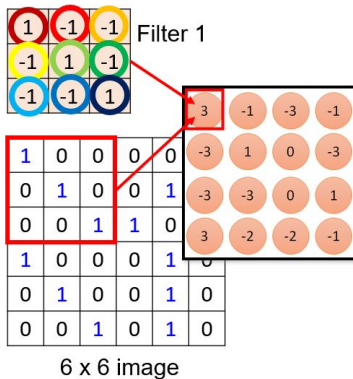
## Convolution vs. Fully Connected NN

Fully-  
connected

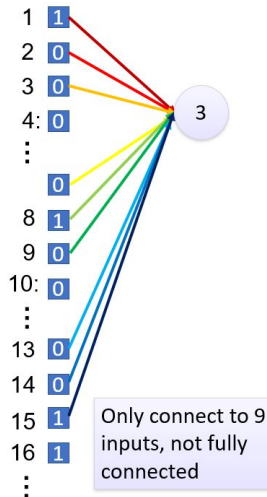
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
1	0	0	0	1	0
0	1	0	0	1	0
0	0	1	0	1	0



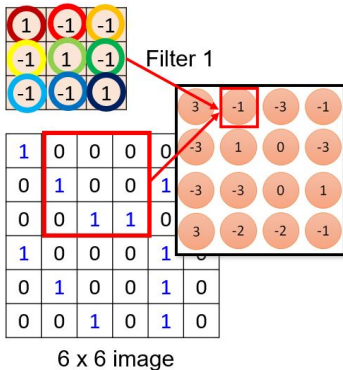
# Convolution vs. Fully Connected NN



fewer parameters!

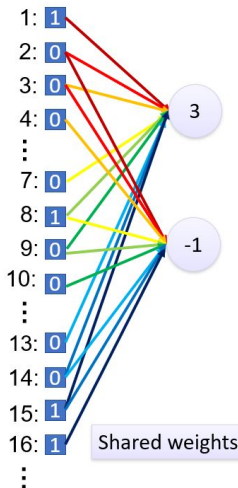


# Convolution vs. Fully Connected NN



Fewer parameters

Even fewer parameters



# Pooling

- In convolutional neural networks (CNNs), pooling is a down-sampling operation commonly used to reduce the spatial dimensions of the input volume.
- The most common type of pooling is called max pooling and average pooling.

Pool size ← → Stride →

-4	0	-2	4	1
3	1	0	2	1
1	0	1	1	1
4	6	5	1	0
-1	2	0	0	0

**Max Pooling**

3	4
6	5

**Min Pooling**

-4	-2
-1	0

**Average Pooling**

0	1
2	1

# 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

## Max pooling

# Why pooling

- Subsampling pixels will not change the object
- We can subsample the pixels to make image smaller
- fewer parameters to characterize the image

bird



Subsampling

bird





# Max pooling

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

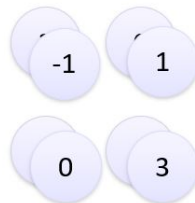
## Max pooling

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



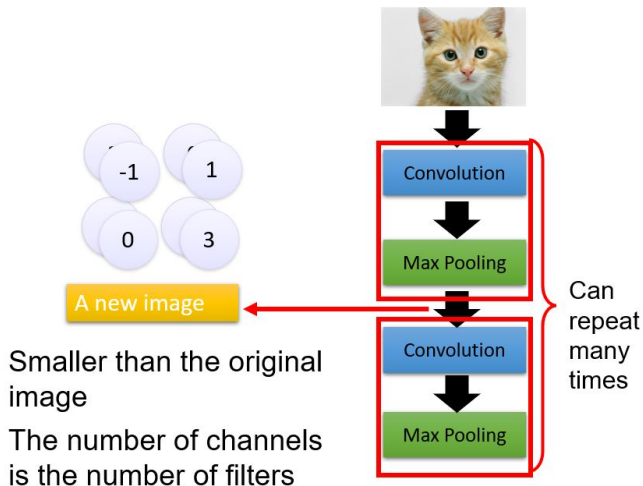
Conv

Max  
PoolingNew image  
but smaller

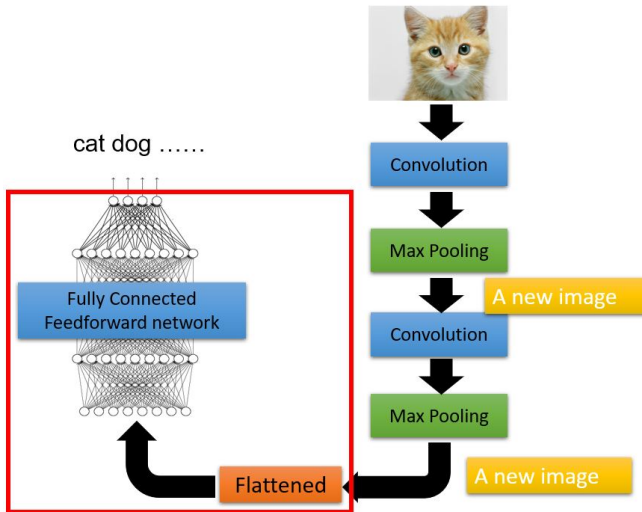
2 x 2 image

Each filter  
is a channel

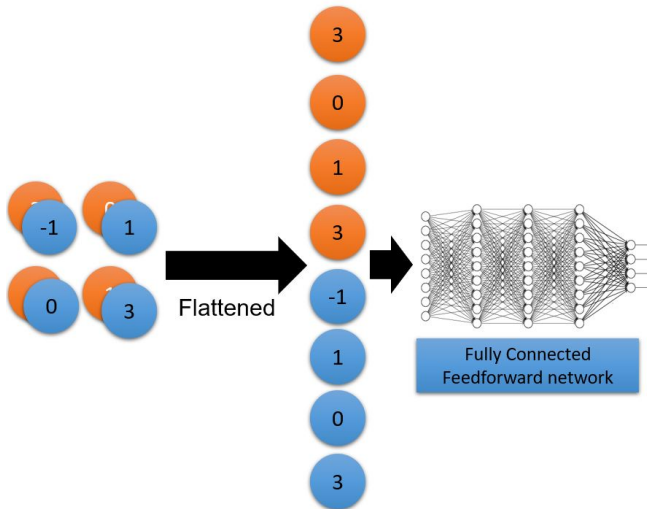
# The whole CNN



# The whole CNN



# Flattening



## CNN in Keras

```
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

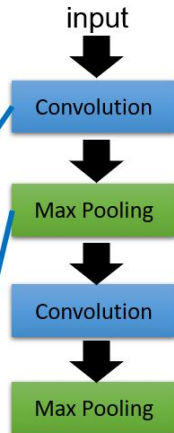
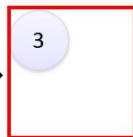
There are  
**25 3x3**  
filters.

Input\_shape = ( 28 , 28 , 1)

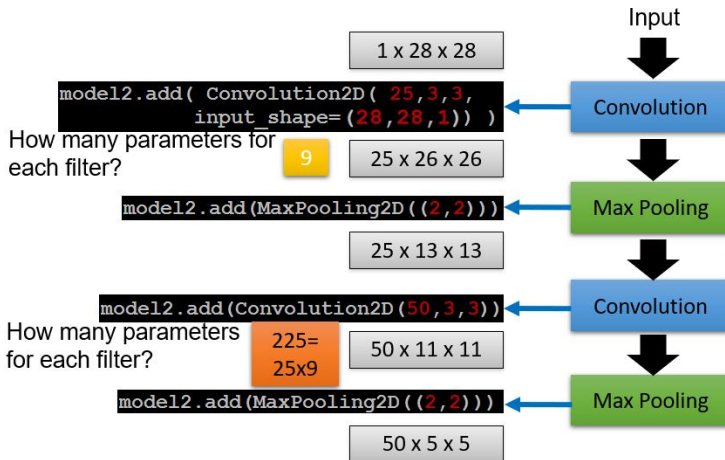
28 x 28 pixels

1: black/white, 3: RGB

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



## CNN in Keras



## CNN in Keras

