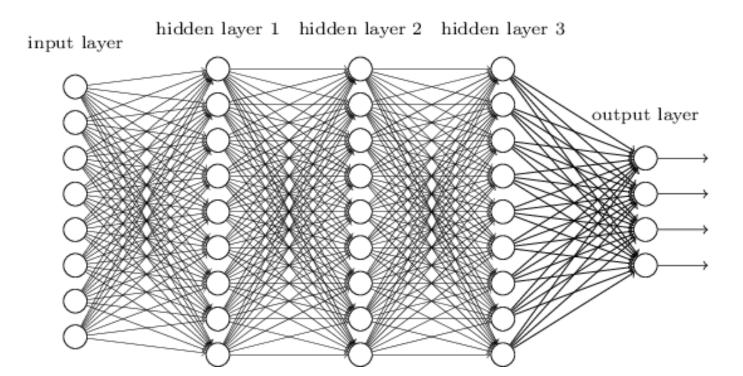
## Lecture 5 Smaller Network: CNN

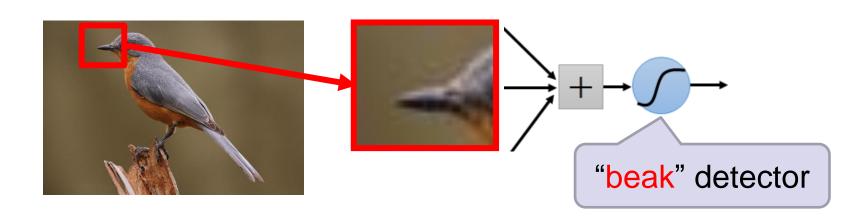
- We know it is good to learn a small model.
- From this fully connected model, do we really need all the edges?
- Can some of these be shared?



# Consider learning an image:

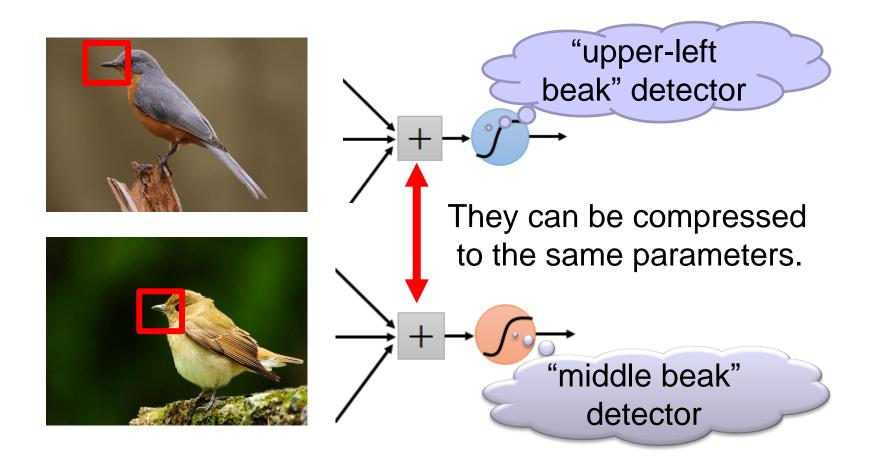
 Some patterns are much smaller than the whole image

Can represent a small region with fewer parameters



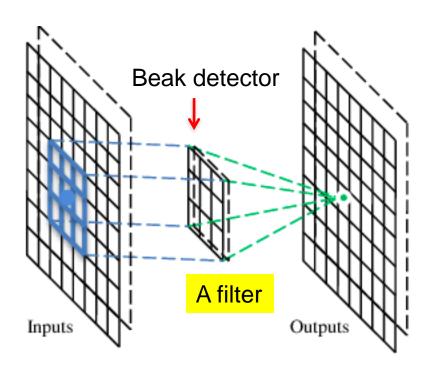
Same pattern appears in different places: They can be compressed!

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.



1	0	0	0	0	1
0	~	0	0	1	0
0	0	~	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
Υ_	~	<b>-</b>
-1	-1	1

Filter 1



Filter 2

: :

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

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 3

3 ) ( -1

6 x 6 image

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

3

-3

-1 -1 -1 -1 \ -1 -1 \

Filter 1

#### stride=1

1	N.	0	0	0	0	1
	0		0	0	1	0
	0	0		1	0	0
	4	$\cap$	Λ	$\cap$	4	Λ
ı		U	0	0		0
	0		0	0	1	0

6 x 6 image



-3 (-3 (0 ) (1

3 -2 -2 -1

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

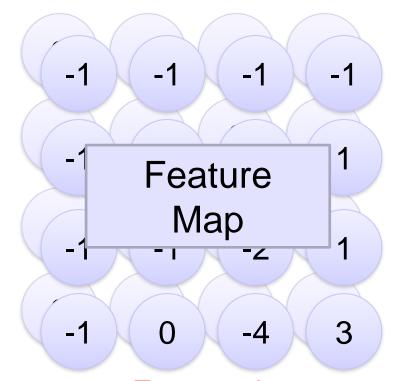
Filter 2

#### 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

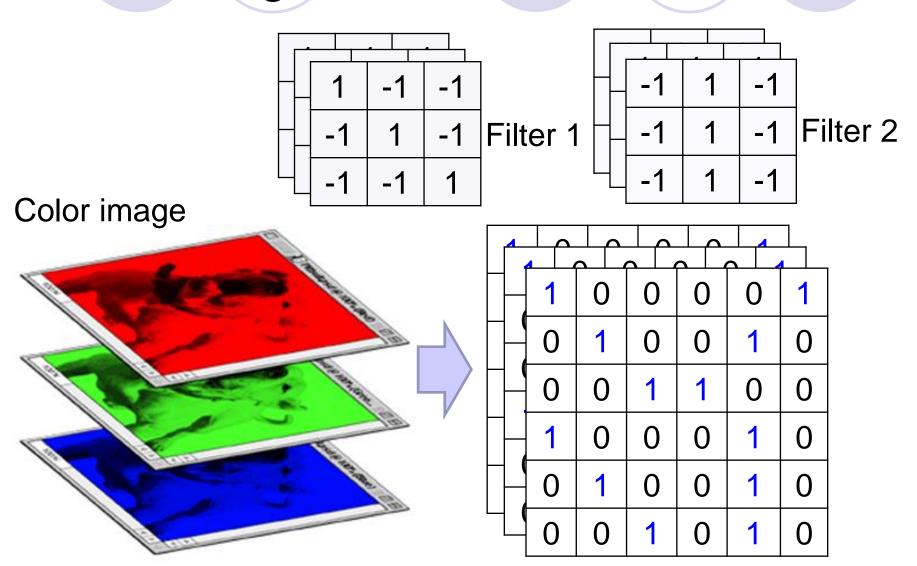
6 x 6 image

#### Repeat this for each filter

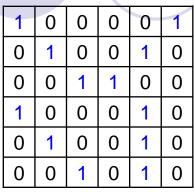


Two 4 x 4 images
Forming 2 x 4 x 4 matrix

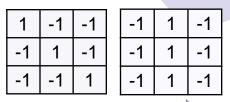
# Color image: RGB 3 channels



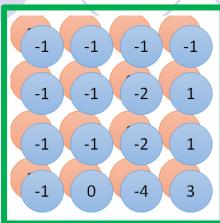
#### Convolution v.s. Fully Connected



image

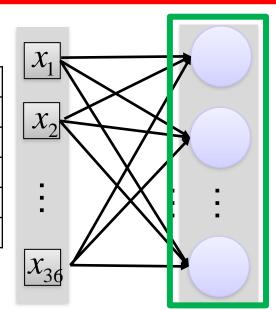


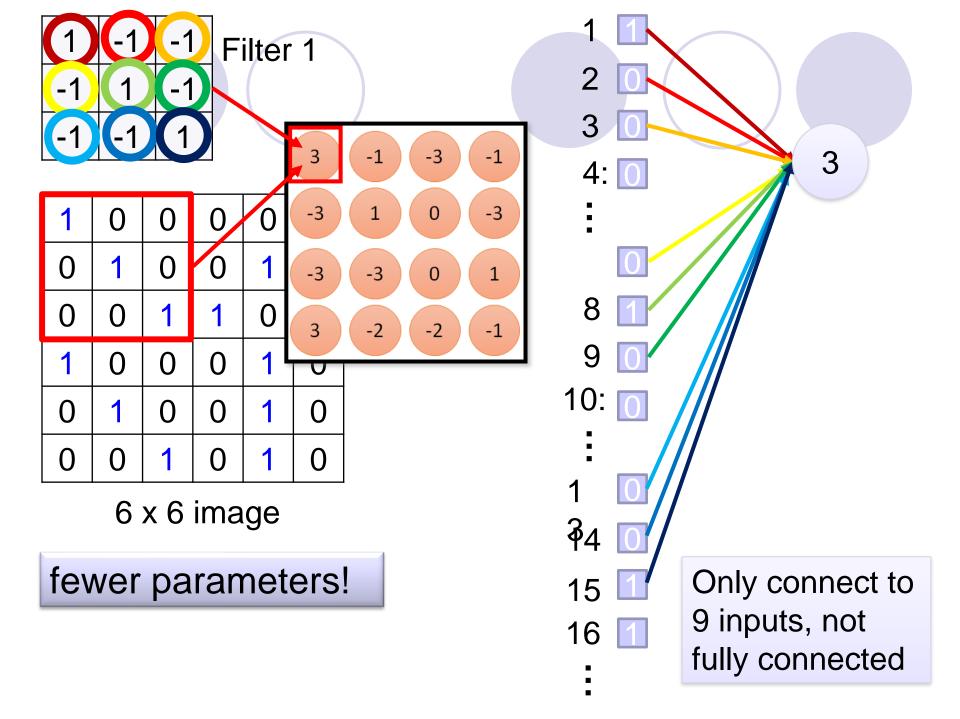
convolution

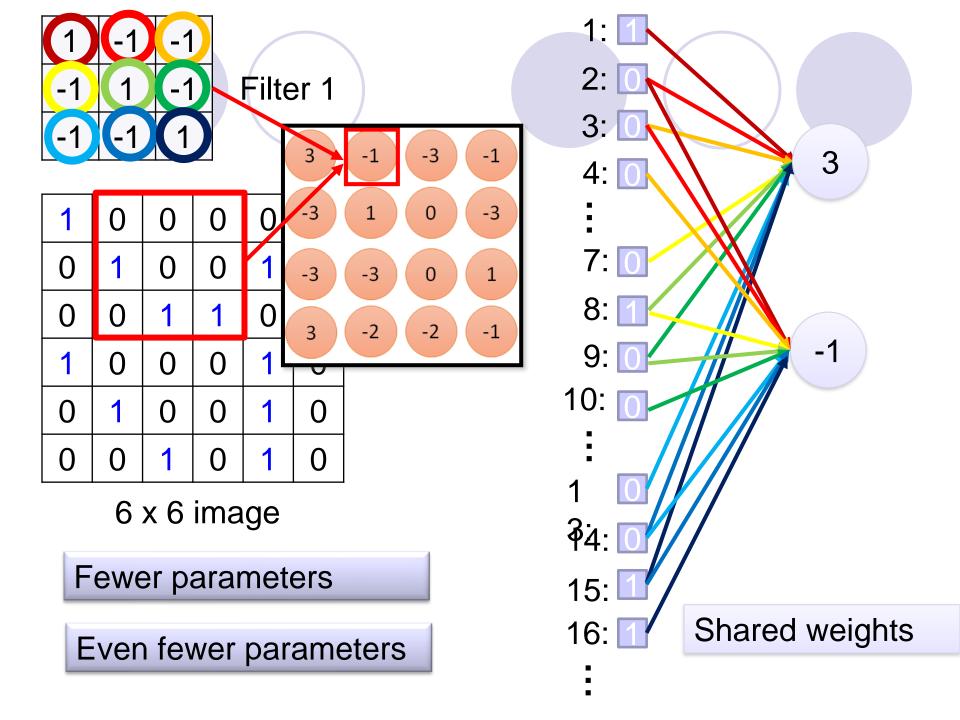


Fullyconnected

1	0	0	0	0	1
0	1	0	0	1	0
0	0	~	1	0	0
1	0	0	0	1	0.
0	1	0	0	1	0:
0	0	1	0	1	0

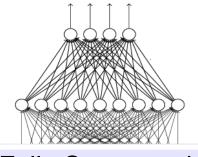




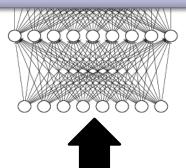


# The whole CNN

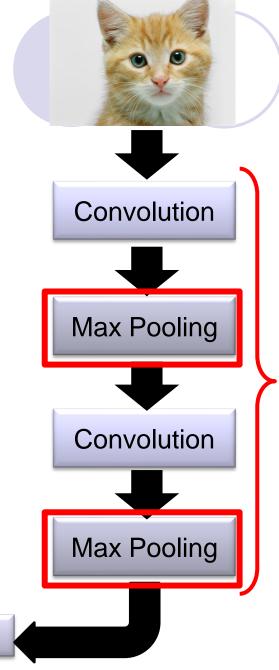
cat dog .....



Fully Connected Feedforward network



Flattened

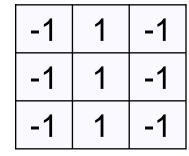


Can repeat many times

# Max Pooling

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

Filter 1



Filter 2

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

3 -2

-3

0	1
-2	-1

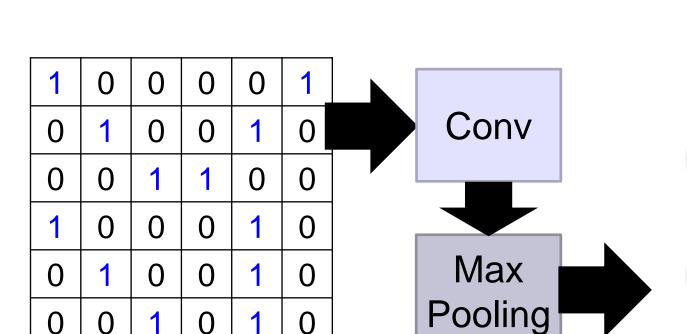
-1
-1
-1
-1

_			-2
_	1	1	-2
_	1	0	-4

# 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



6 x 6 image

# New image but smaller

-1 1

0 3

2 x 2 image

Each filter is a channel

### The whole CNN

-1 1

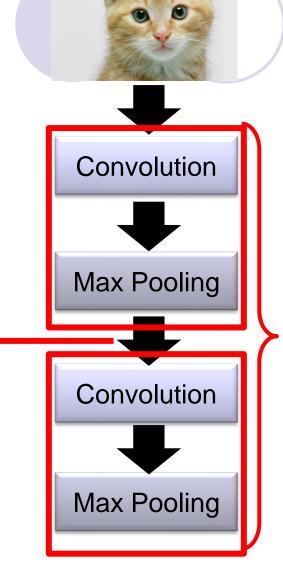
3

A new image

0

Smaller than the original image

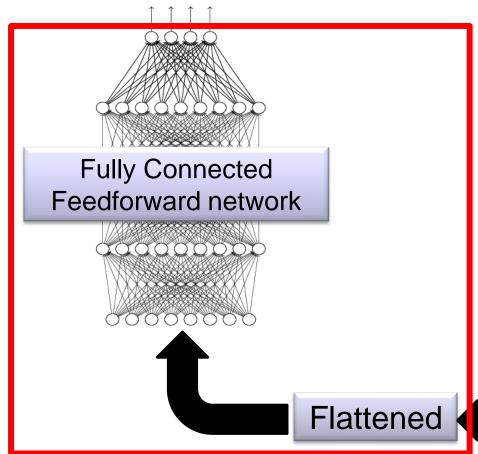
The number of channels is the number of filters

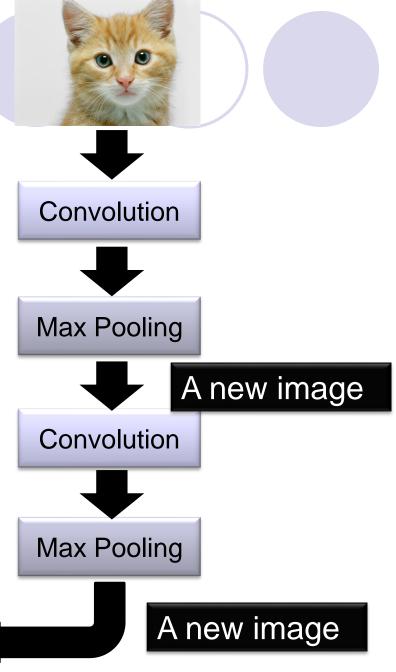


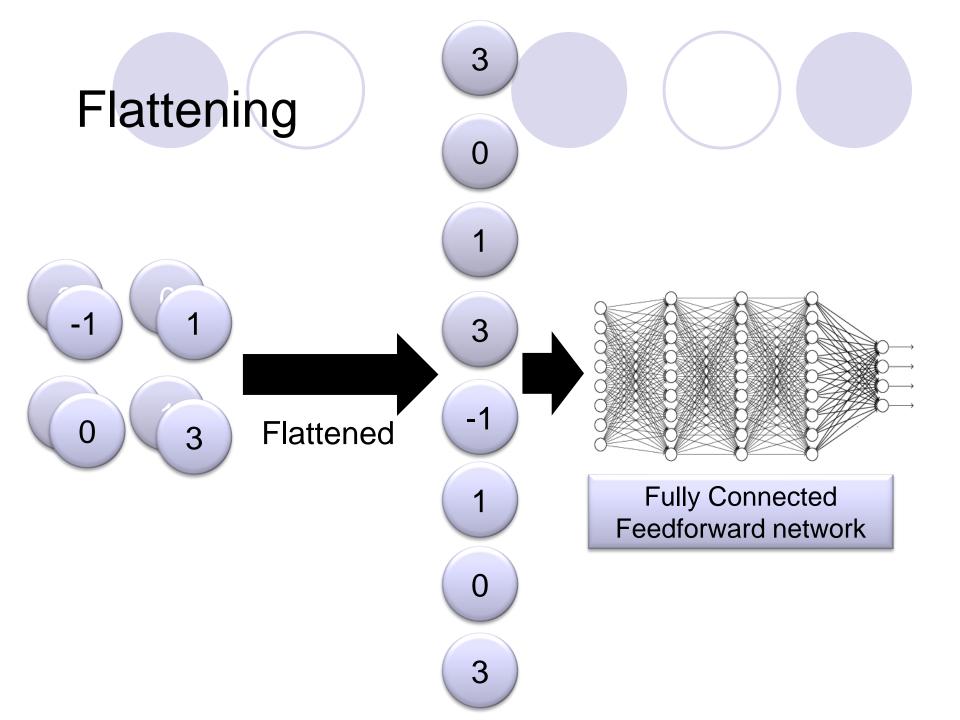
Can repeat many times

## The whole CNN

cat dog .....

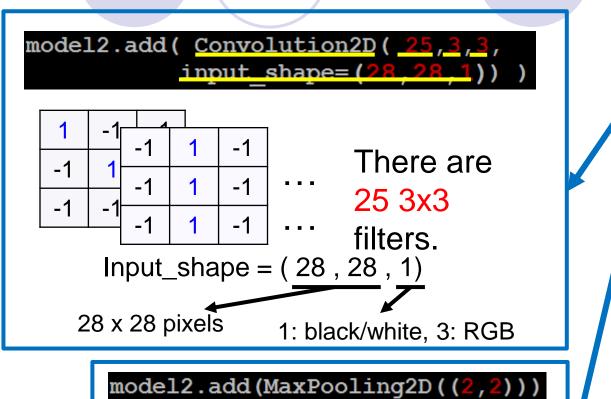


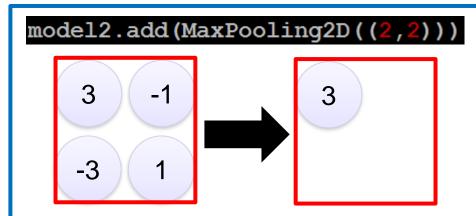


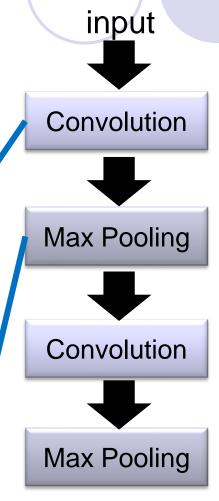


#### CNN in Keras

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

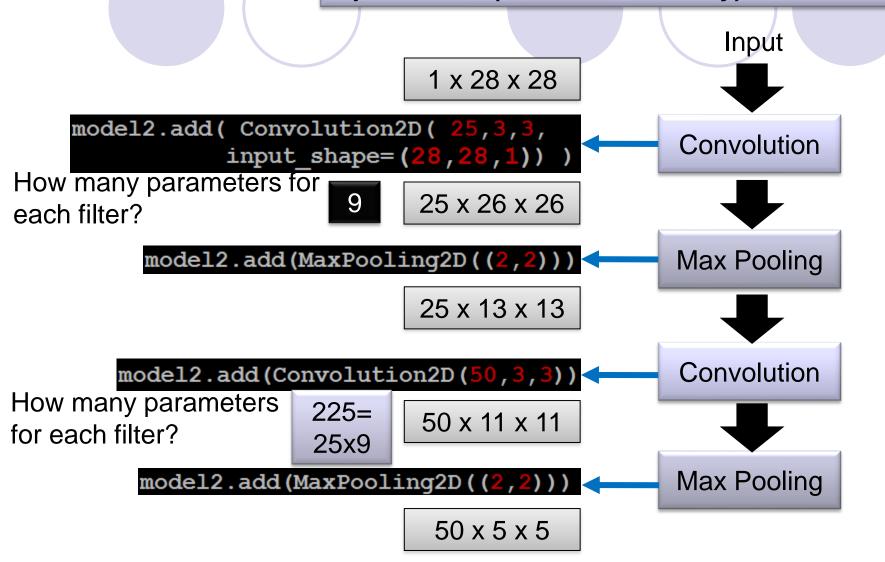






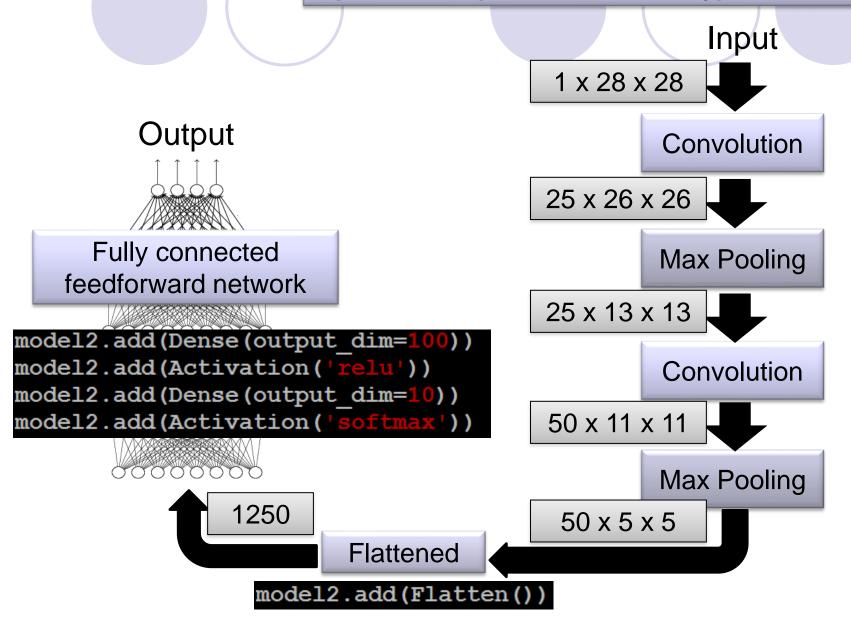
#### CNN in Keras

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)* 



```
import tensorflow as tf
def generate model():
   model = tf.keras.Sequential([
       tf.keras.layers.Conv2D(32, filter size=3, activation='relu'),
       tf.keras.layers.MaxPool2D(pool size=2, strides=2),
      tf.keras.layers.Conv2D(64, filter size=3, activation='relu'),
       tf.keras.layers.MaxPool2D(pool size=2, strides=2),
     tf.keras.layers.Flatten(),
     tf.keras.layers.Dense(1024, activation='relu'),
     tf.keras.layers.Dense(10, activation='softmax')
  ])
  return model
                                        CONVOLUTION + RELU
                                                  POOLING
                                                        CONVOLUTION + RELU POOLING
                                                    FEATURE LEARNING
```