

# Convolutional Neural Networks

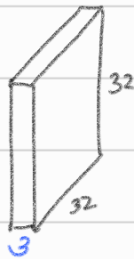
Hierarchical organization 1959 고봉이 뇌 실험

Simple cells — Complex cells — Hypercomplex cells

ImageNet Classification with Deep CNN 2012 "AlexNet"

## Convolutional Layer

ex.  $32 \times 32 \times 3$  image : spatial structure 보존



$5 \times 5 \times 3$  filter  $w$



Convolve the filter  
with the image  
(단장적합이)

Filters always extend the full depth of the input volume : 3

number : the result of taking a dot product btw  
the filter and a small  $5 \times 5 \times 3$  chunk of the img.

$$w^T x + b$$

모든 spatial location에 convolve 하면

$28 \times 28 \times 1$  activation map 생성

또 다른 filter로 다른 activation map 생성

예를 6개 maps 만들면  $28 \times 28 \times 6$

ConvNet 은 Convolution Layers 의 Sequence, interspersed with activation f.

ex. Conv Layer - RELU - Conv Layer - RELU - POOL - ... - "FC"

Input img size :  $N$

Filter size :  $F$

일 때, Output map 은  $(N - F) / \text{stride} + 1$

$$(32 - 5) / 1 + 1 = 28$$

## Zero Padding (in practice)

pad with 1 pixel border

$\therefore$  maintain the same input size

filter size  $\neq$   $3 \times 3$  : zero pad with 1

$5 \times 3$  : " 2

$7 \times 3$  : " 3

Input volume :  $32 \times 32 \times 3$

10  $5 \times 5$  filters w/ stride 1, pad 2  $\frac{1}{2}$  then,

$$\# \text{ of params} = (\underbrace{5 \times 5 \times 3}_{\text{bias}} + 1) \times 10 = 760$$

## Pooling layer : downsample

- makes the representations smaller and more manageable
- operates over each activation map independently
- depth is the same

## Max Pooling

1	1	2	4
5	6	7	8
3	2	1	0
1	2	3	4



6	8
3	4

max pooled with  
 $2 \times 2$  filters and stride 2