

# CNN

한양대학교 인공지능연구실 인공지능세미나

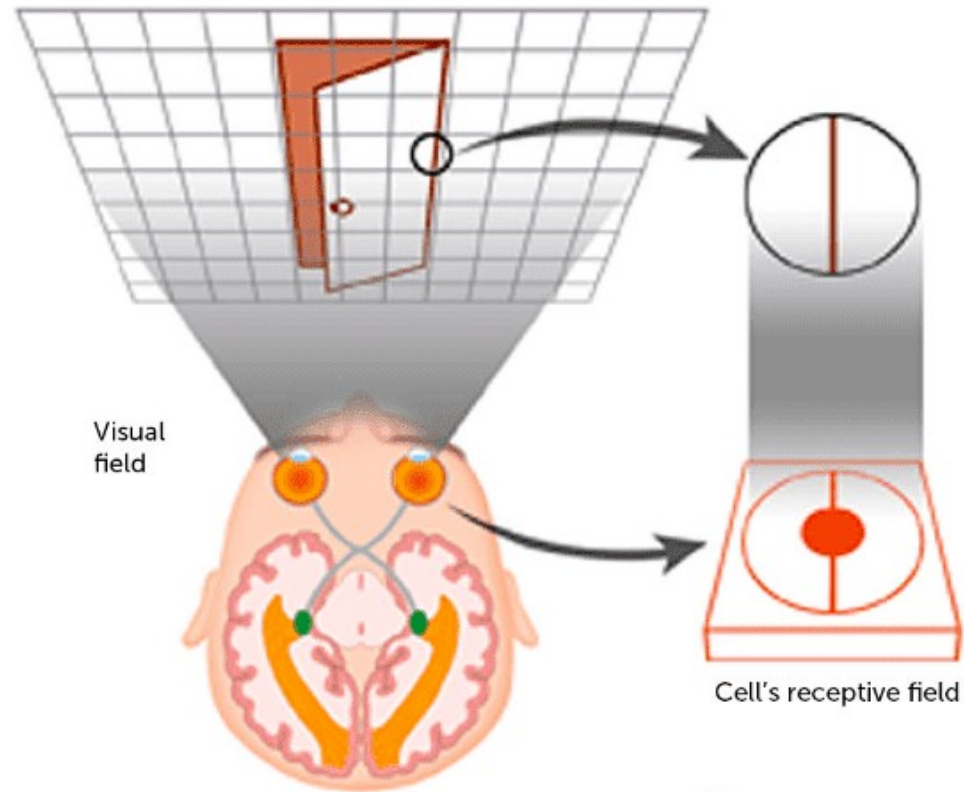
20220126  
AILAB 조환희

## **Convolutional** Neural Network

# Convolutional?

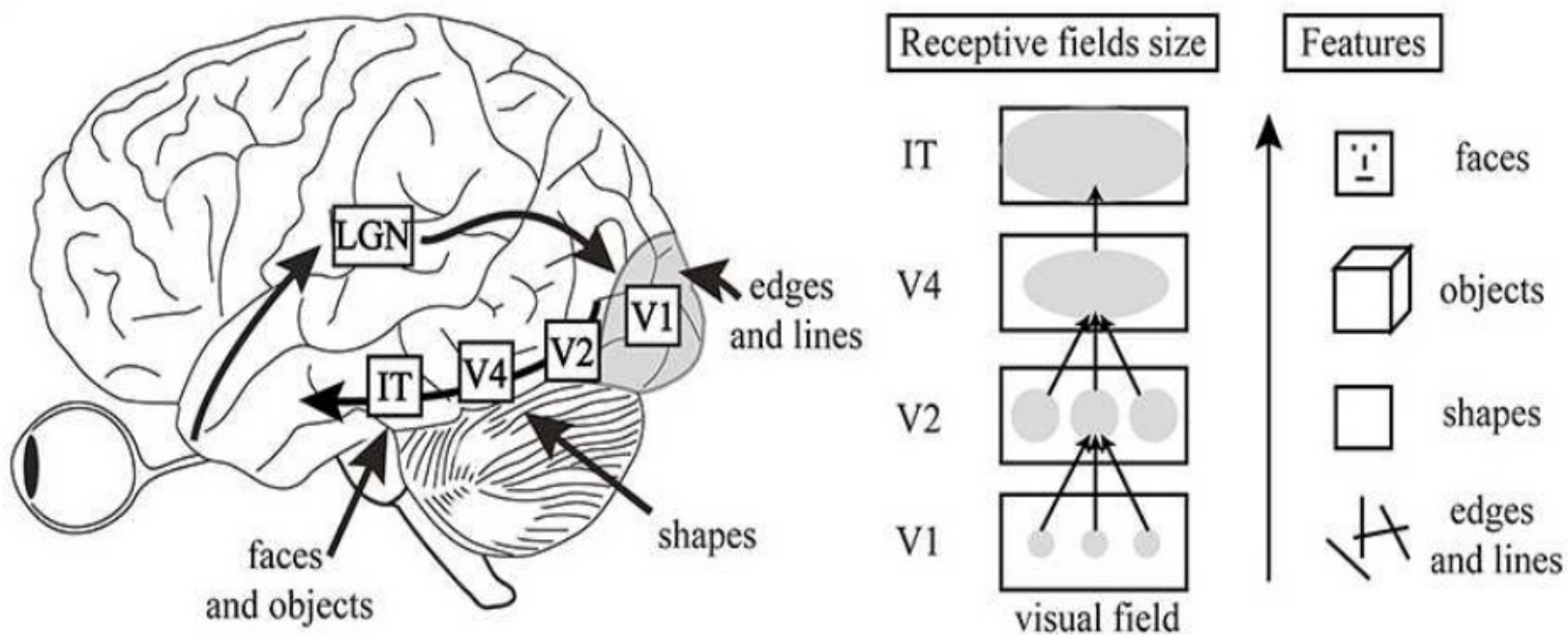
## Receptive Field

한 뉴런이 수용하는 이미지 ( 감각 ) 정보 크기



# Convolutional?

하나의 이미지를 여러 성분을 조합하여 생성한다 !



# Convolutional?



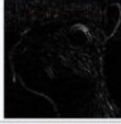
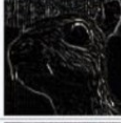

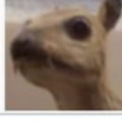
앞선 성질을 Deep Learning 에서 적용을 어떻게 할까 ?

## --> Convolution

이미지 각 픽셀에 대해 필터의 값들을 곱한후 합치는 과정

== 필터의 weight 를 조작해서

이미지에서 원하는 정보를 가져올 수 있다 !

Operation	Kernel	Image result
Identity	$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$	
Edge detection	$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}$	
	$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$	
	$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$	
Sharpen	$\begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{bmatrix}$	
Box blur (normalized)	$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$	

# Convolutional?

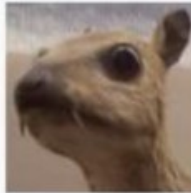

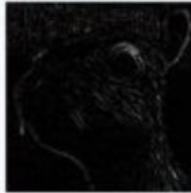
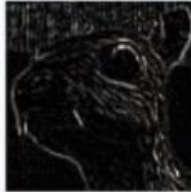
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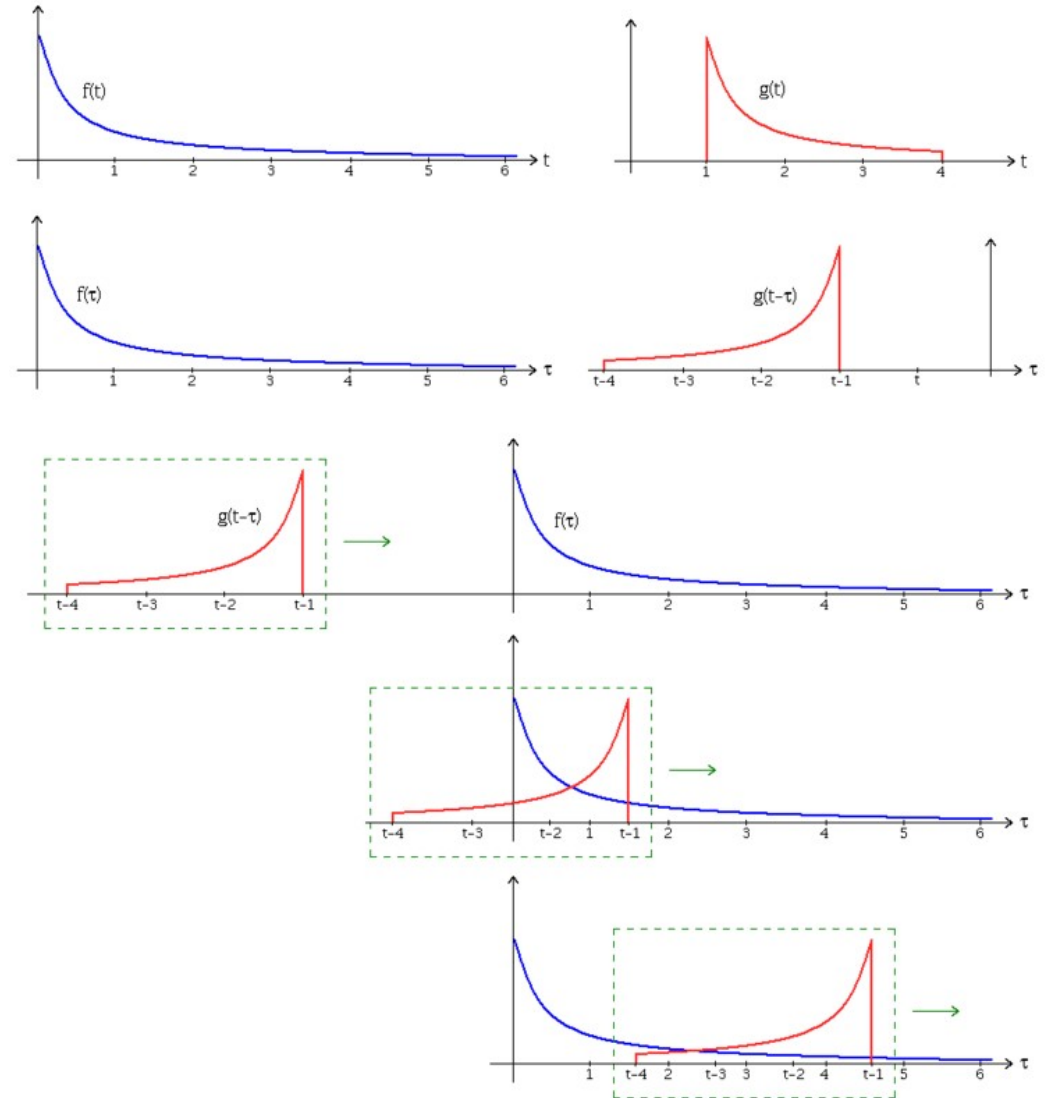
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	$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$	

# Convolutional?

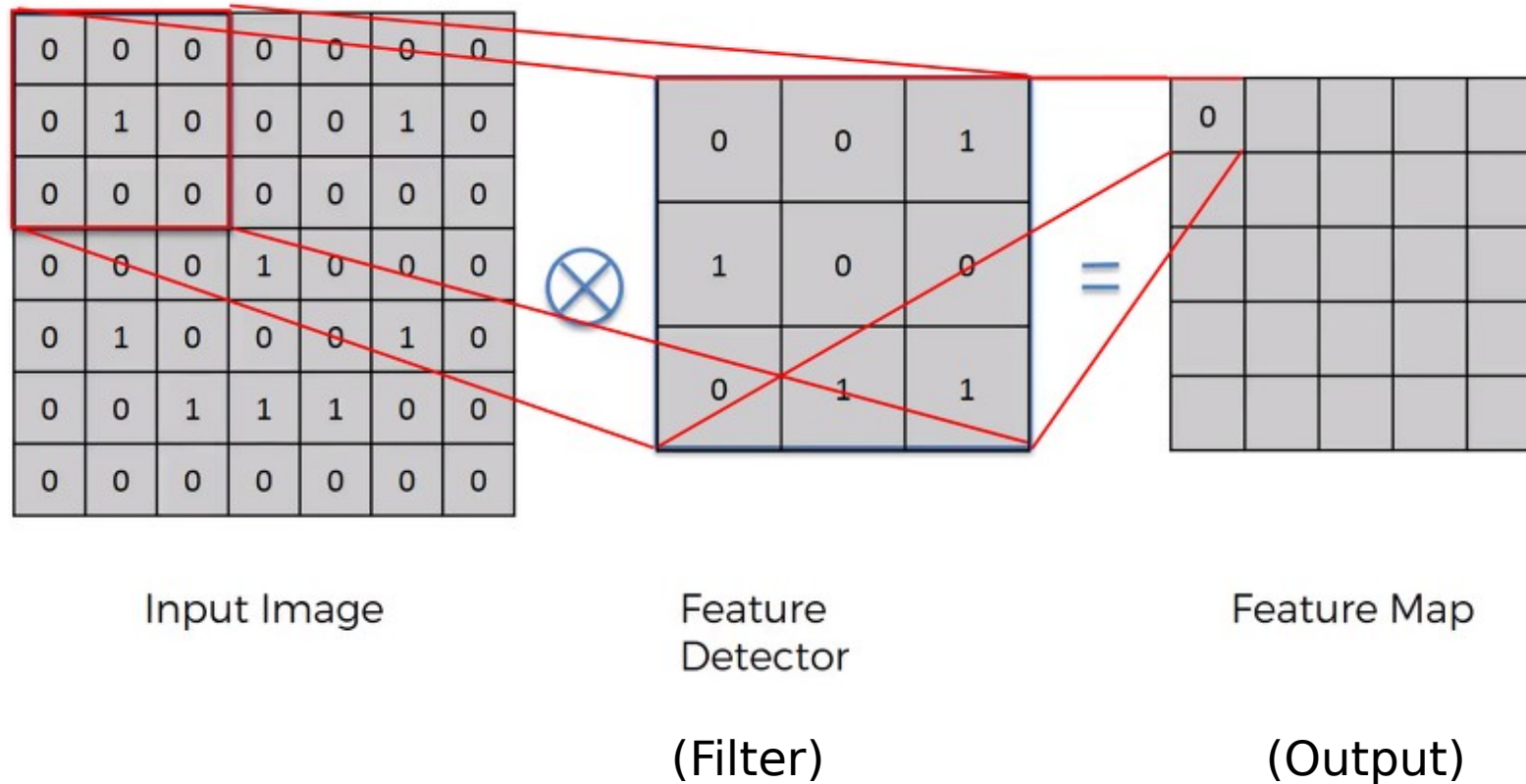
## Convolution

$$(f * g)(t) \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau$$



# Convolutional?

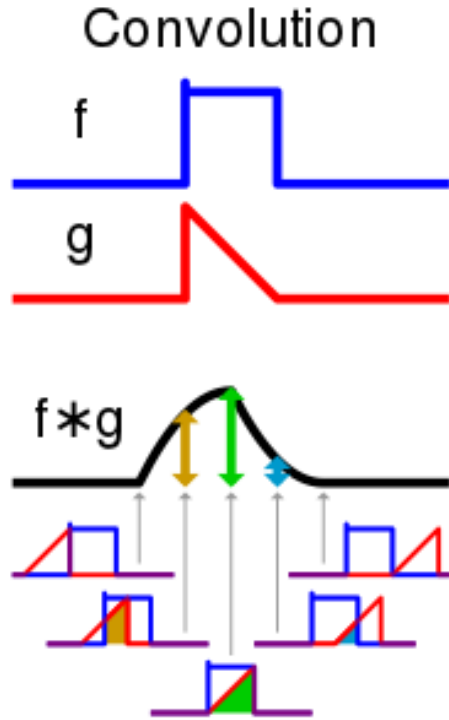
## Convolution in Deep Learning





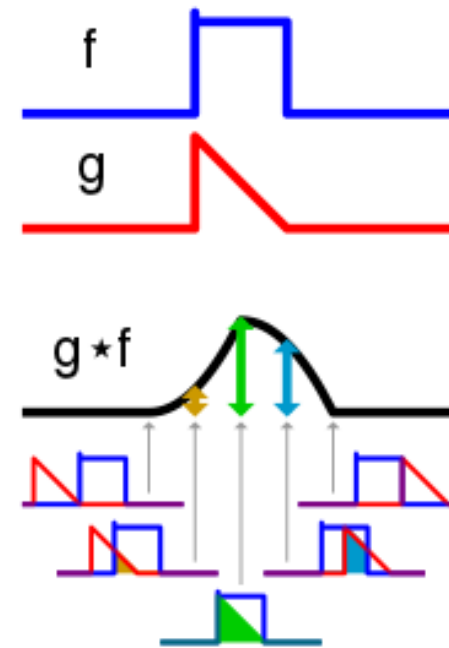
# Convolutional?

## Convolution



$$(f * g)(t) \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau$$

## Cross-correlation



$$(f * g)(t) \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} f(\tau) g(t + \tau) d\tau$$

# Convolutional?

## Convolution

실제 구현은 cross-correlation

Convolution

image			kernel		
1	2	3	A	B	C
4	5	6	D	E	F
7	8	9	G	H	I

\*

$$(1 * I) + (2 * H) + (3 * G) + (4 * F) + \dots + (9 * A)$$

Cross-correlation

image			kernel		
1	2	3	A	B	C
4	5	6	D	E	F
7	8	9	G	H	I

⊗

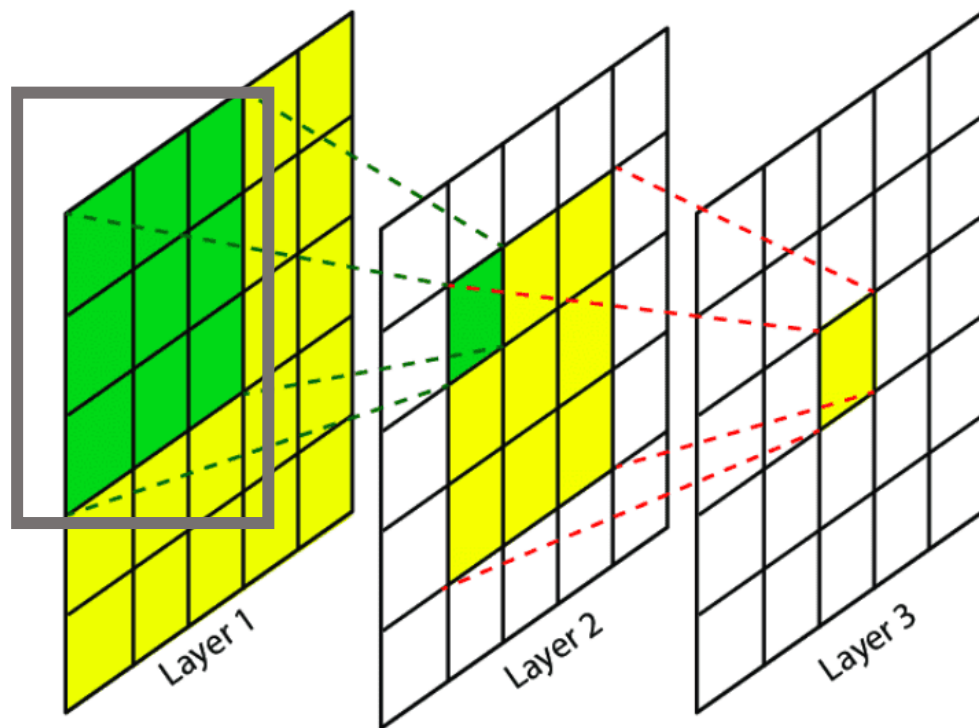
$$(1 * A) + (2 * B) + (3 * C) + (4 * D) + \dots + (9 * I)$$

# Convolutional?

## Receptive Field

한 뉴런이 수용하는 이미지 ( 감각 ) 정보 크기

--> 어떤 feature 를 가져오는 영역의 크기



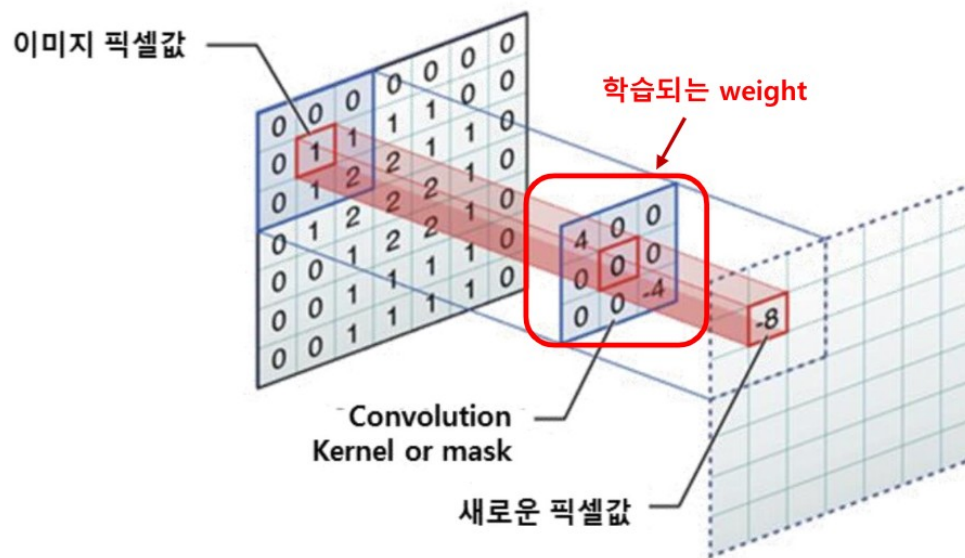
# Convolutional?

## Convolution

Convolution 과 CNN Convolution의 차이

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사람이 설계



신경망이 설계해줌

# Convolutional Neural Network

## LeNet (1998)

1. CNN layer
2. Pooling layer
3. FC layer

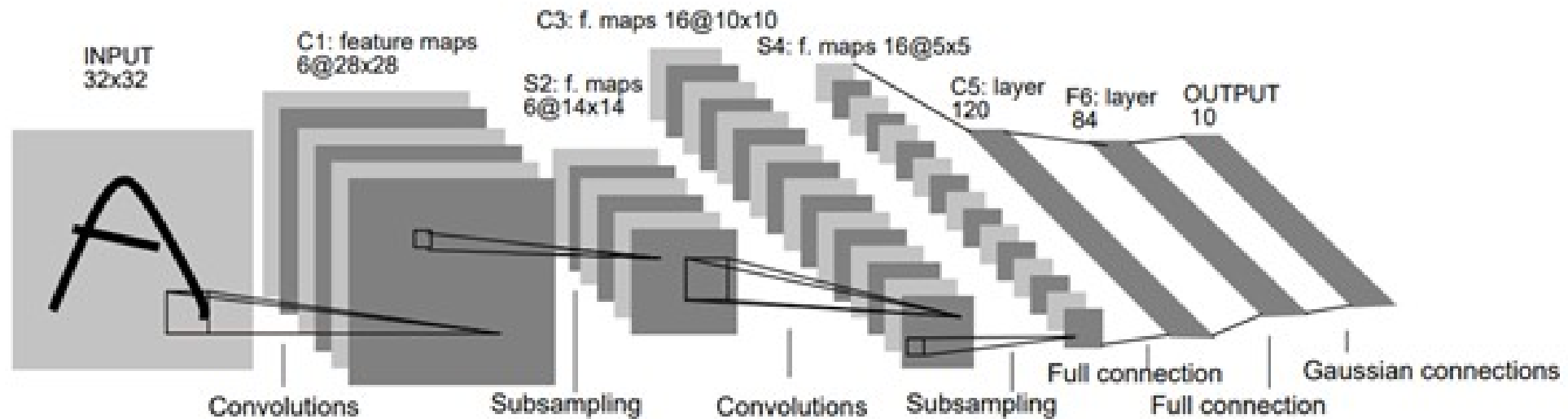
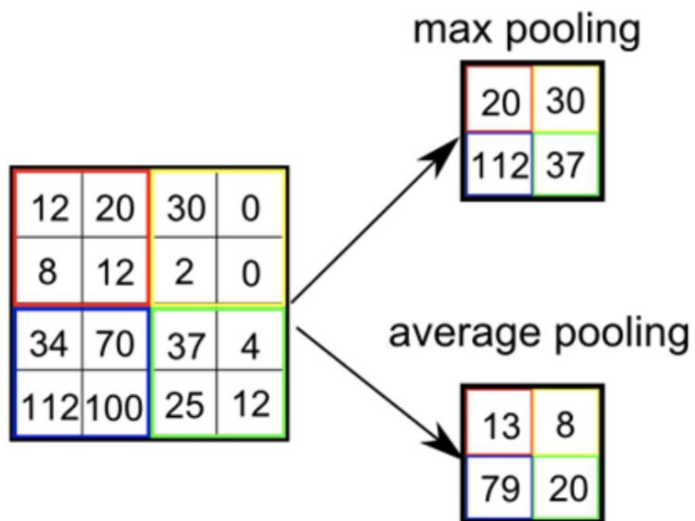


Fig. 2. Architecture of LeNet-5, a Convolutional Neural Network, here for digits recognition. Each plane is a feature map, i.e. a set of units whose weights are constrained to be identical.

# Convolutional Neural Network

## Pooling layer(Sub-sampling/Down-sampling)

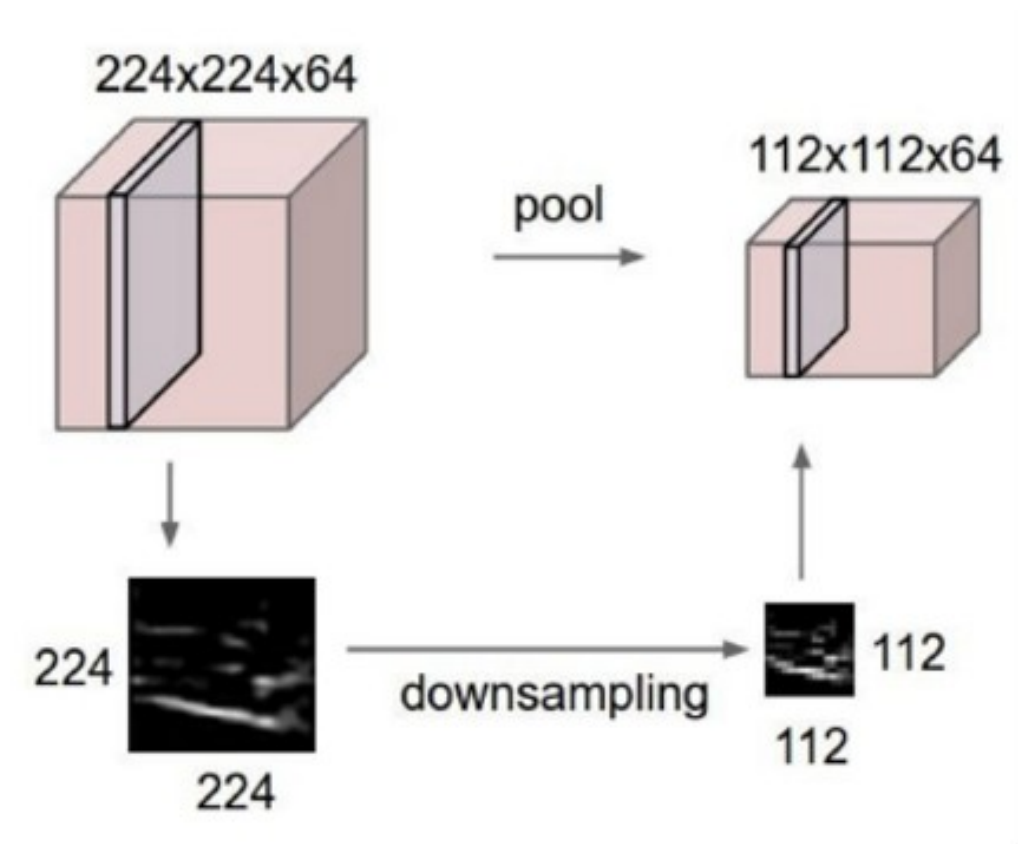


Feature map 사이즈 줄이기  
Strong, Global feature 뽑아 중요한 invariance 얻기

# Convolutional Neural Network

Pooling layer(Sub-sampling/Down-sampling)

정보는 유지된다 !



# Convolutional Neural Network

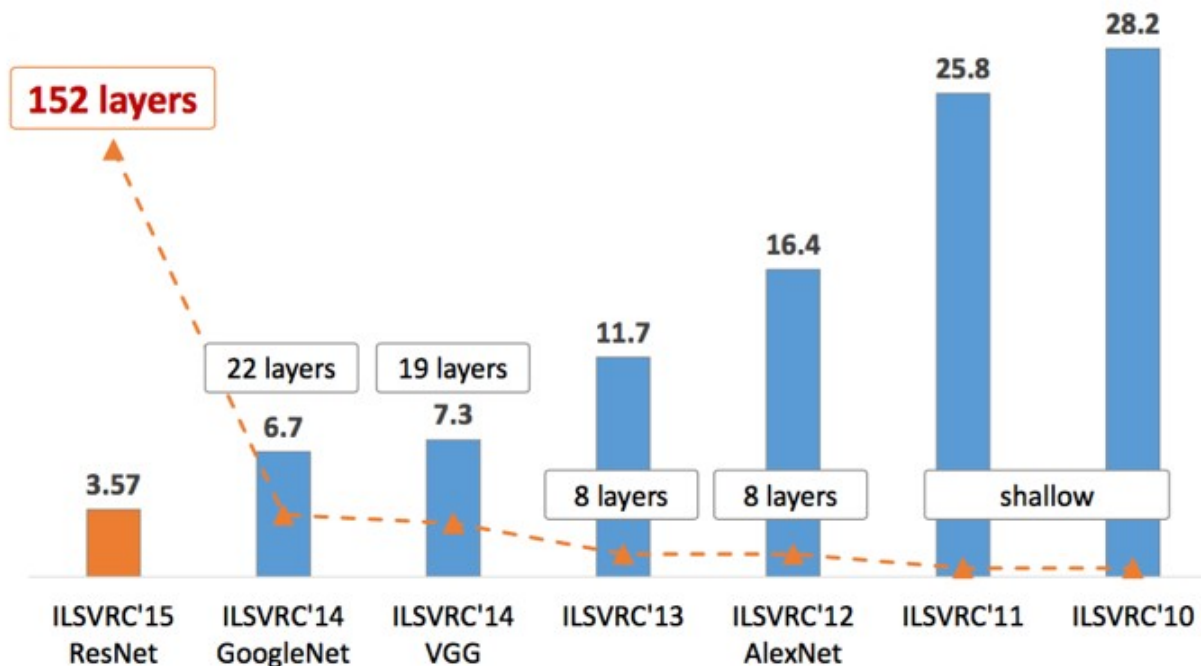
## AlexNet (2012)

### LeNet 구조와 비슷

인풋 사이즈 (32x32 -> 224x224)

모델 사이즈 키움, dropout 사용

CNN 구조가 재조명됨





# Convolutional Neural Network

## ConvNetJS

### CNN 학습과정 / Feature Map 보여주는 사이트

