

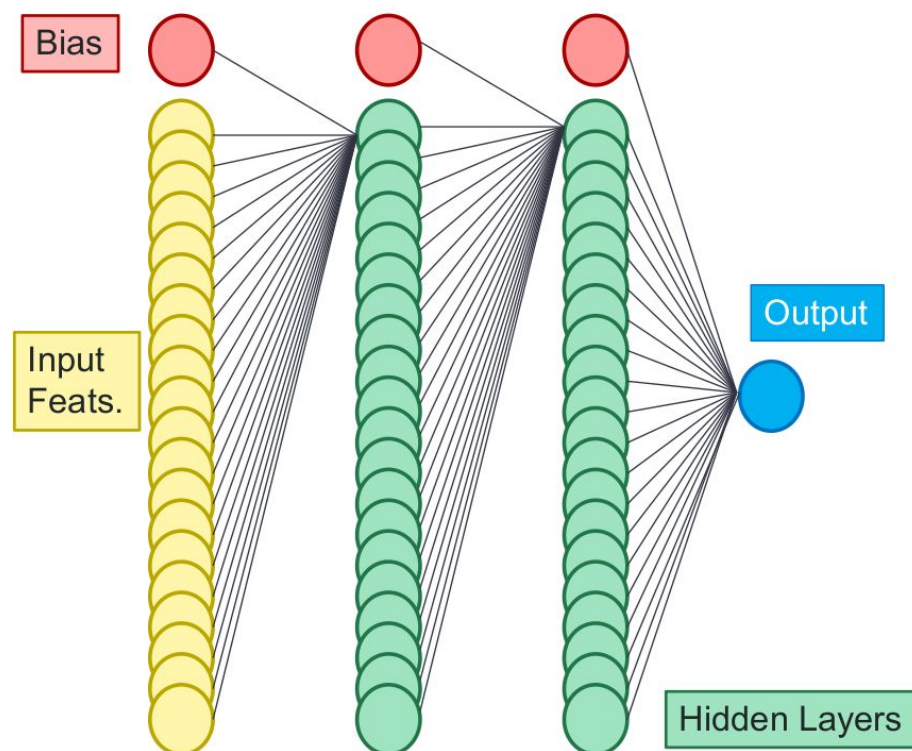
# Introduction to Convolutional Neural Networks

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## Smarter Architectures

An MLP can have MANY parameters



Data: 20 input features, single binary label

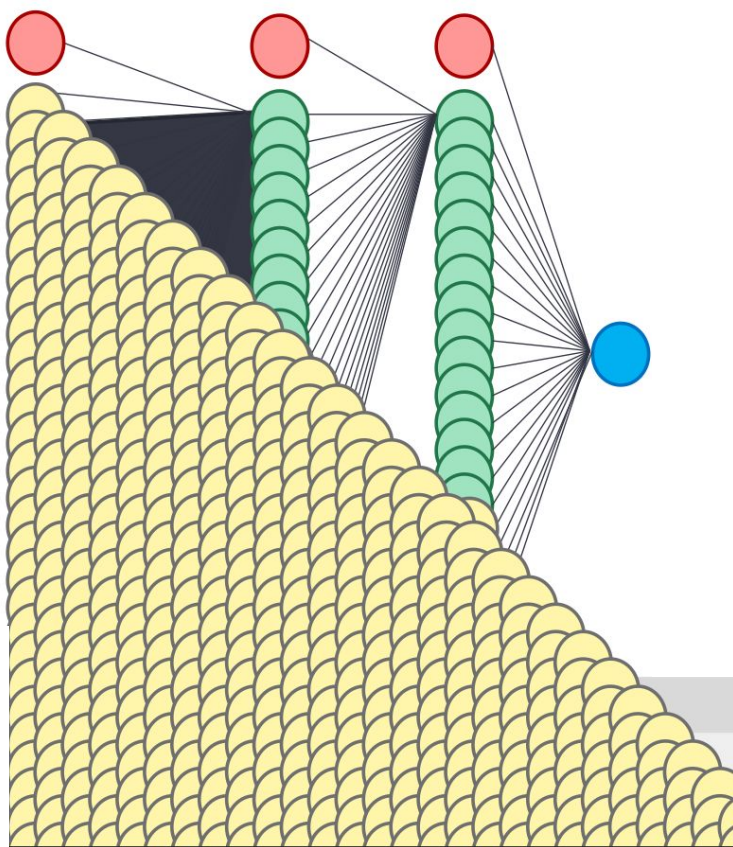
- 1 input layer with 20 nodes,
- 2 fully connected layers, 20 nodes each
- 1 final prediction node

How many weights is that?

$(20+1)*20+(20+1)*20+(20+1)*1=861$   
params

## Smarter Architectures

An MLP can have MANY parameters



Data: 1 input layer with  $256 \times 256$  nodes,  
2 fully connected layers, 20 nodes each  
1 final prediction node  
How many weights is that?  
 $(256 \times 256 + 1) \times 20 + (20 + 1) \times 20 + (20 + 1) \times 1$   
 $= \sim 1.3\text{M params}$

## Smarter Architectures

### ConvNets <3

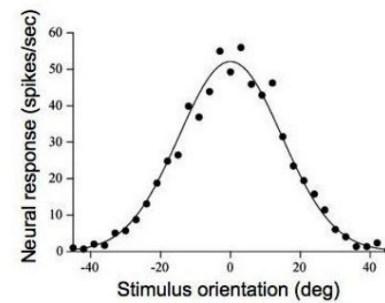
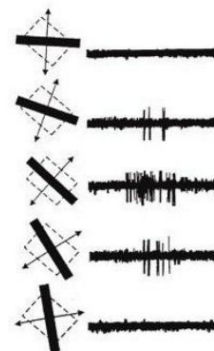
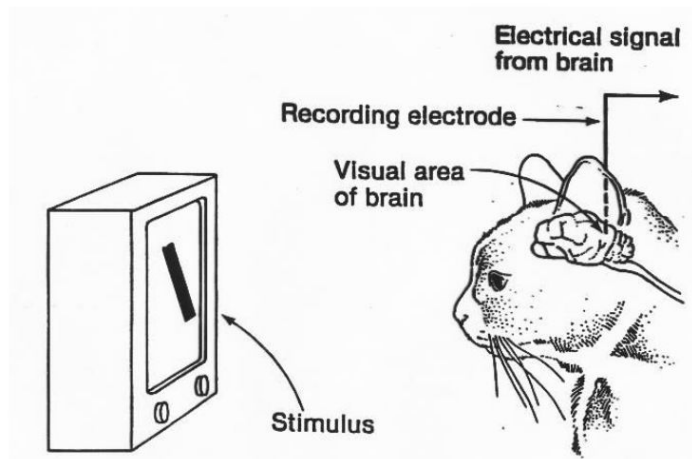
- Convolutional neural networks share parameters across the space -> reduces total number of parameters

## Visual Processing in the Brain

- Brain is an efficient machine -> How it solves vision?

## Visual Processing in the Brain

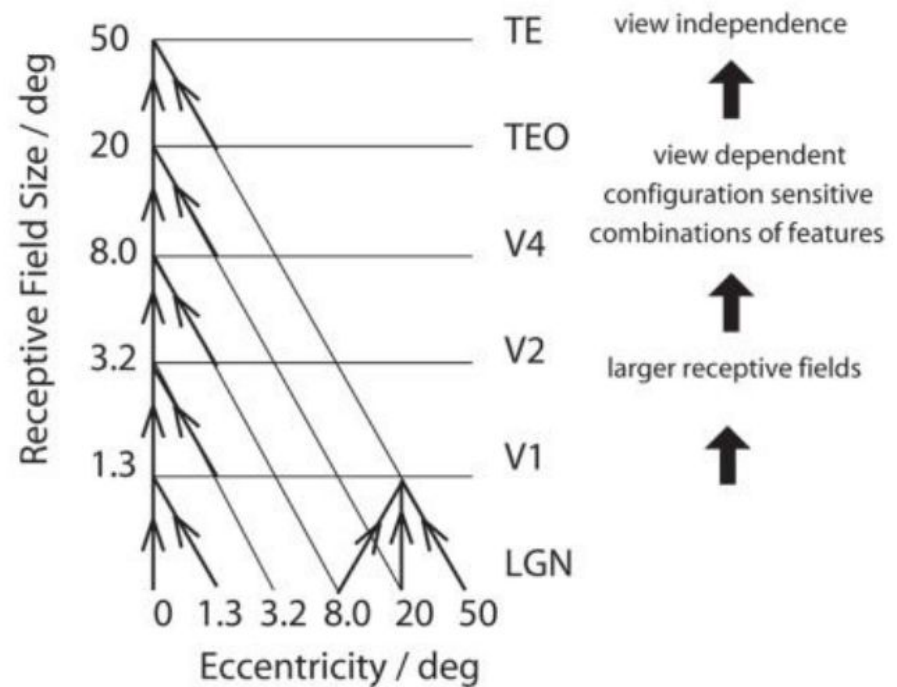
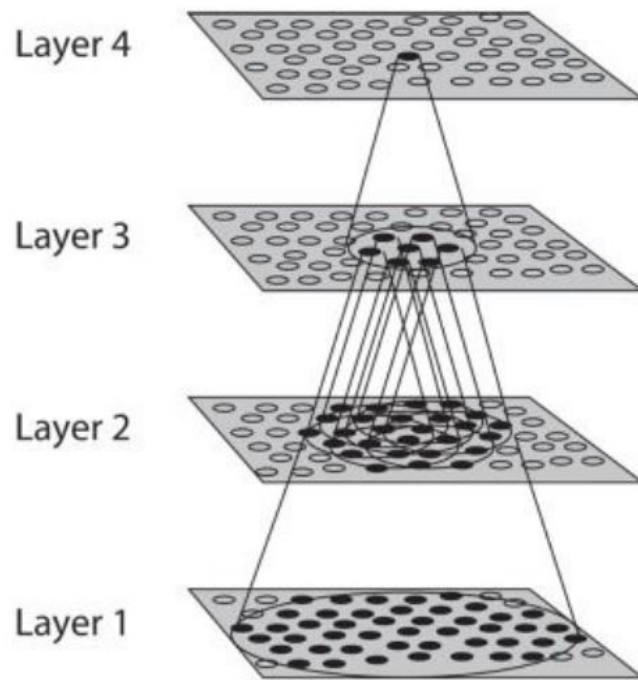
- **Feature selectivity:** Hubel & Wiesel, 1968



Hubel & Wiesel, 1968

## Visual Processing in the Brain

- Hierarchy of processing



## Visual Processing in the Brain

- Invariance

Translation Invariance



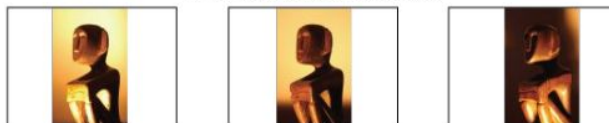
Rotation/Viewpoint Invariance



Size Invariance



Illumination Invariance

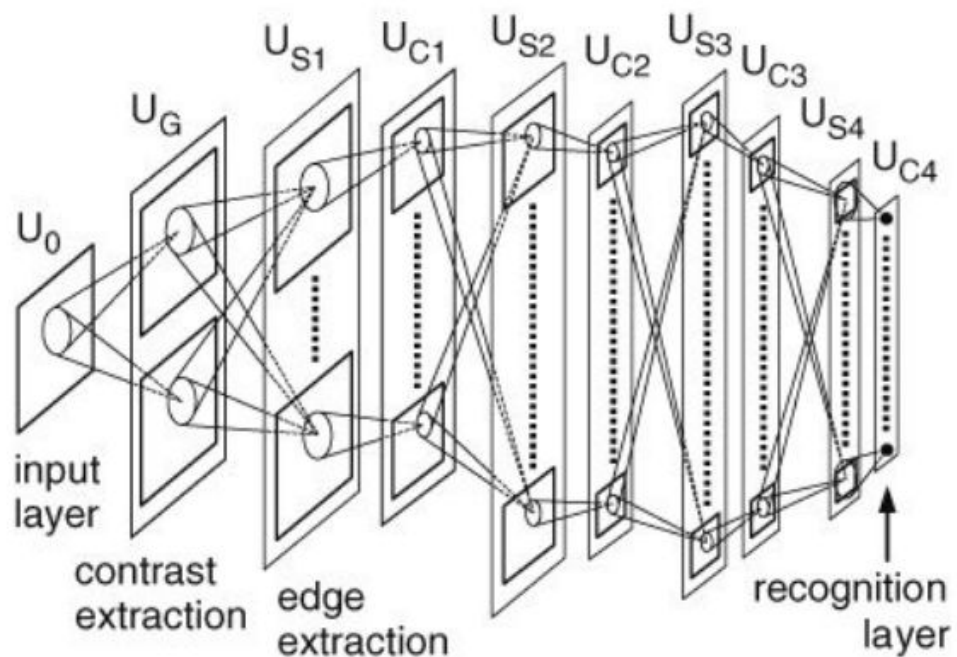


Matt Krause  
mattkrause



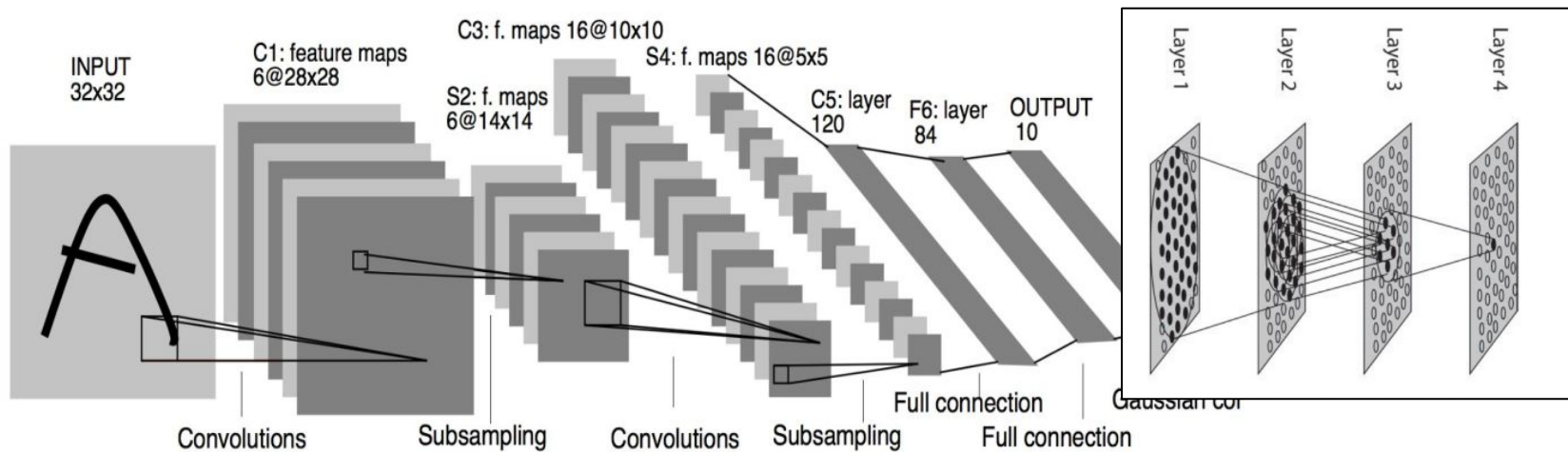
## History of CNNs

Neocognitron (1980) Fukushima



## History of CNNs

LeNet (1998) - Yann LeCun @Bell Labs



Recipe: convolution + subsampling (pooling) + hierarchy

## History of CNNs

CNNs are everywhere!

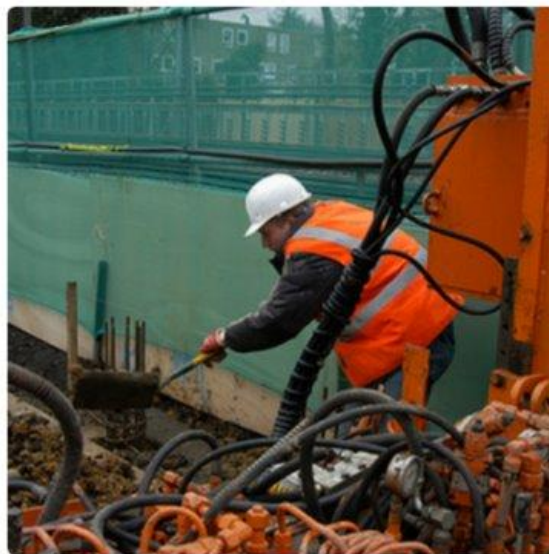


## History of CNNs

CNNs are everywhere!



"man in black shirt is playing guitar."



"construction worker in orange safety vest is working on road."



"two young girls are playing with lego toy."



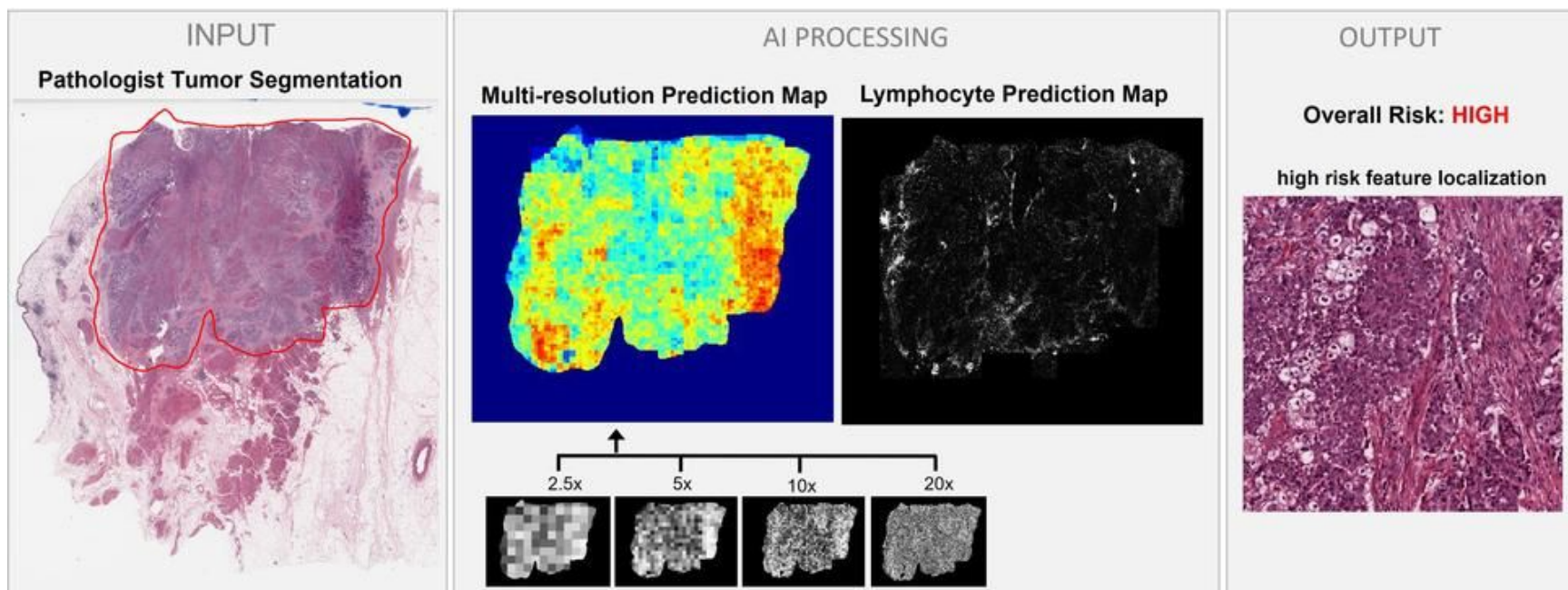
## History of CNNs

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## History of CNNs

CNNs are everywhere!



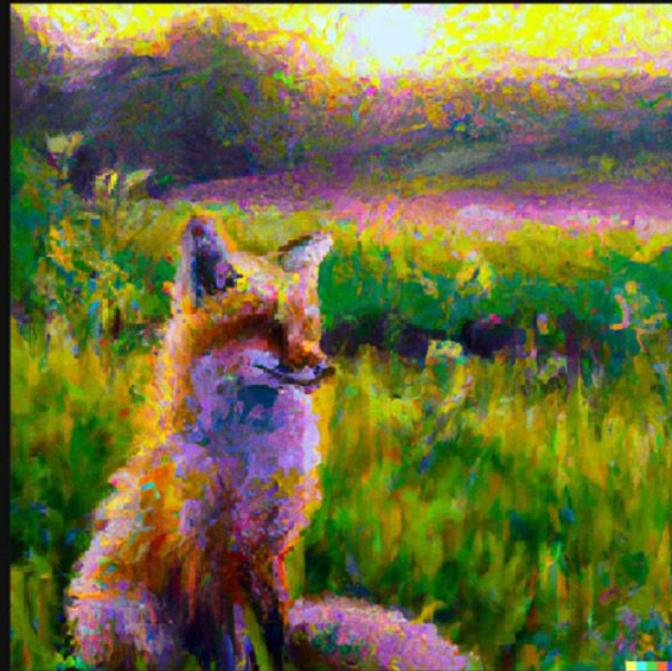
## History of CNNs

CNNs are everywhere!

DALL·E 1



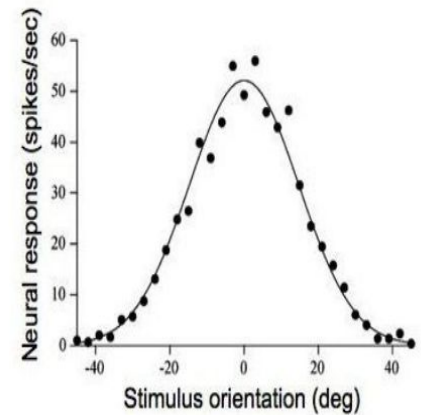
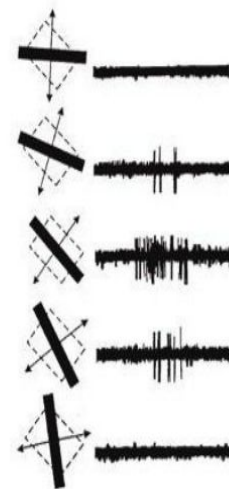
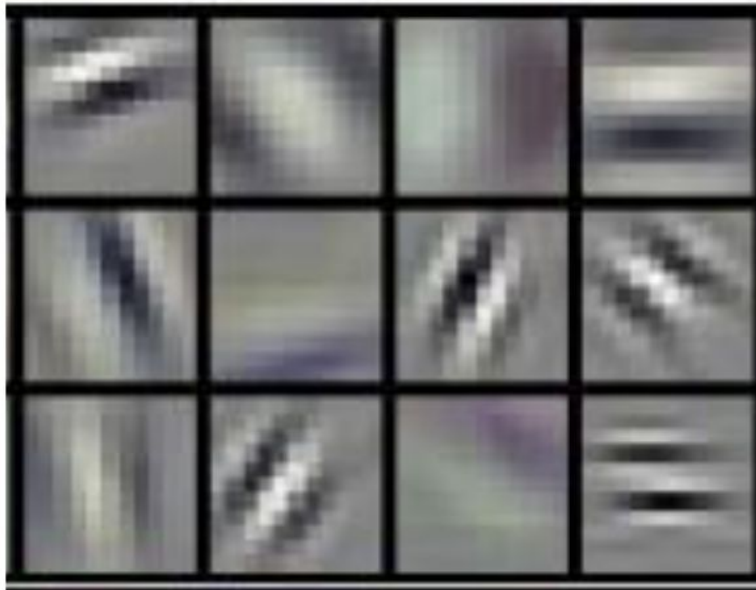
DALL·E 2





## CNN Recipe

Recipe: **convolution** + subsampling (pooling) + hierarchy

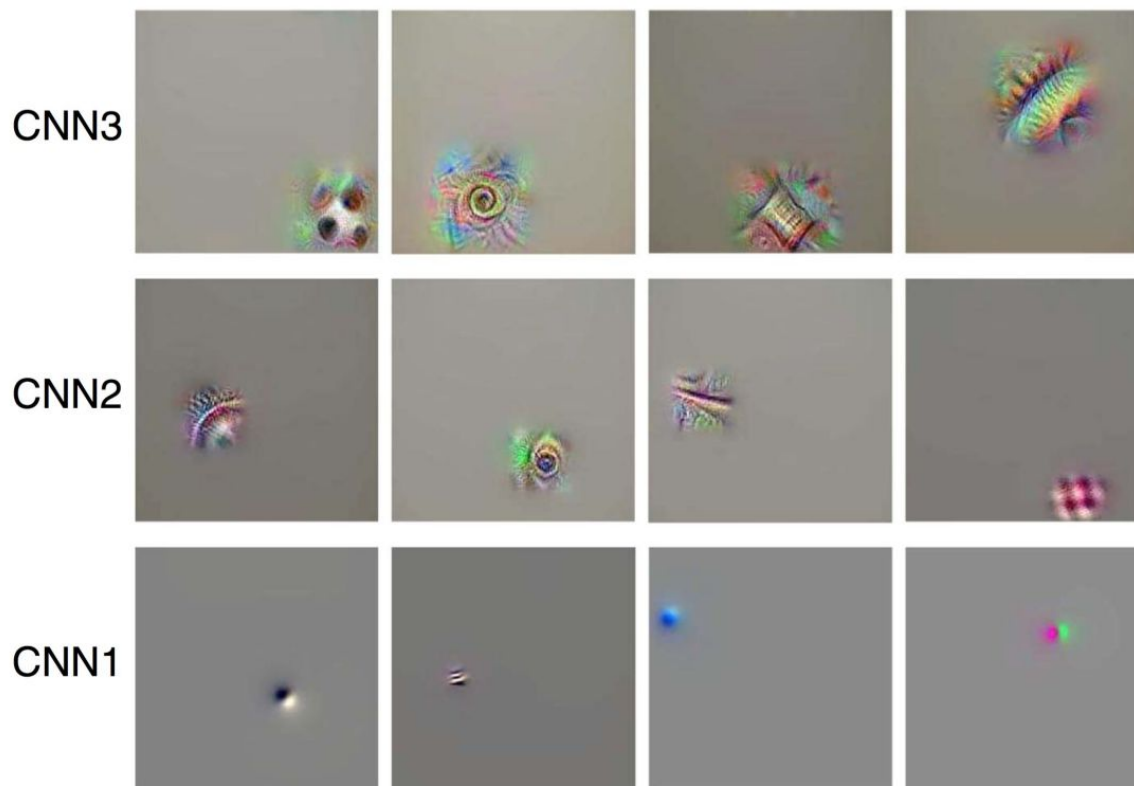


Hubel & Wiesel, 1968



## CNN Recipe

Recipe: **convolution** + subsampling (pooling) + hierarchy



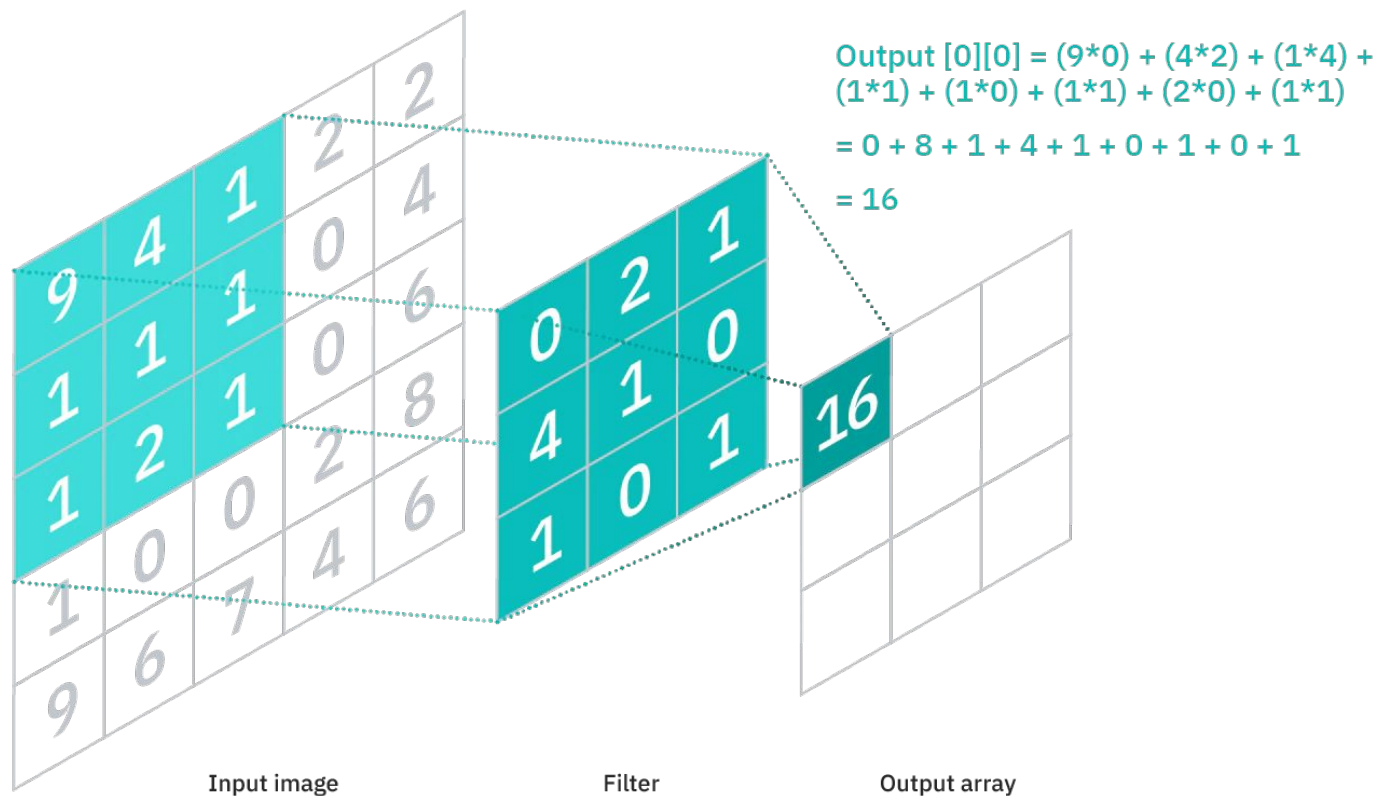
## CNN Recipe

Recipe: **convolution** + subsampling (pooling) + hierarchy



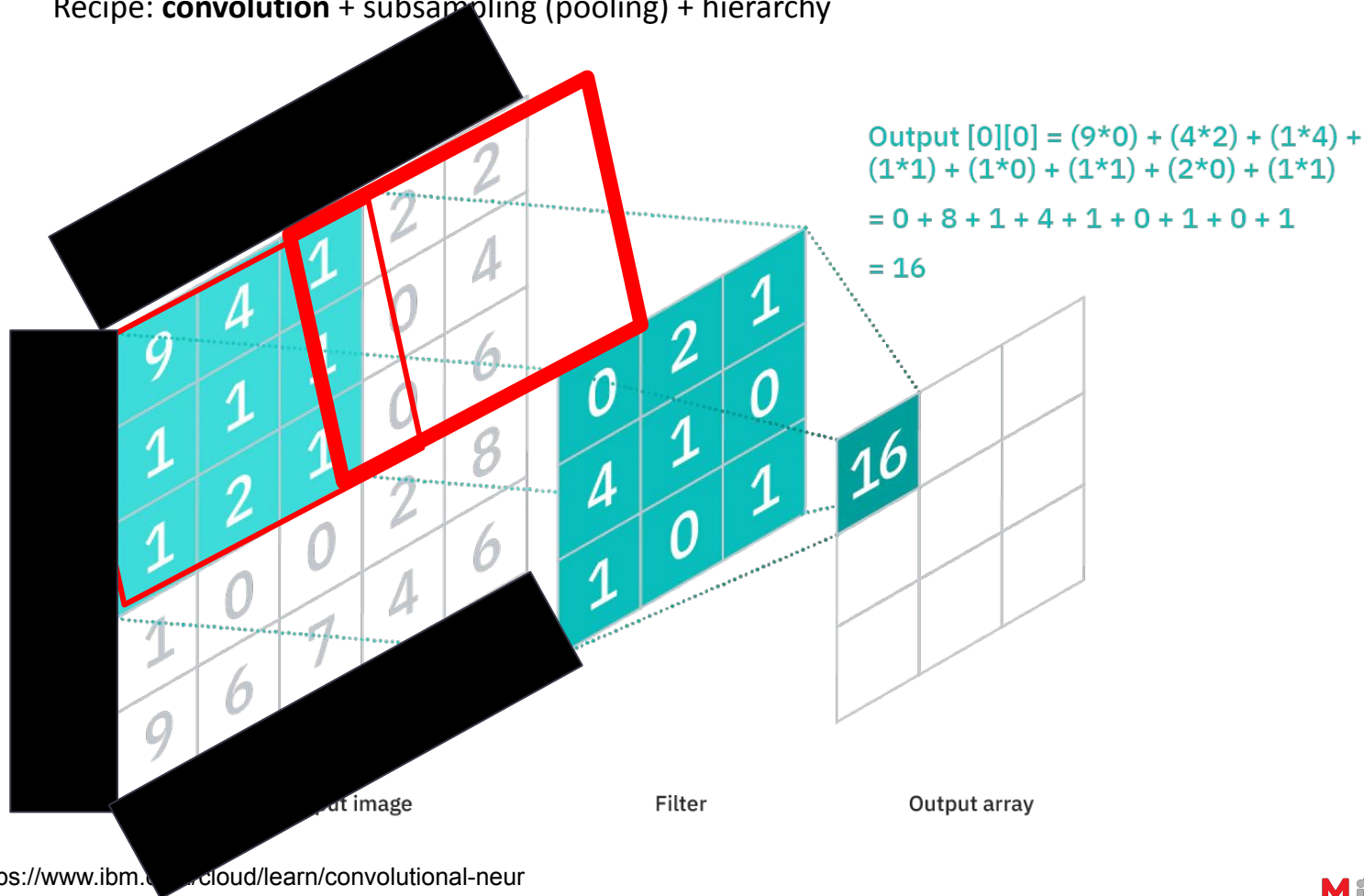
## CNN Recipe

Recipe: **convolution** + subsampling (pooling) + hierarchy



## CNN Recipe

Recipe: **convolution** + subsampling (pooling) + hierarchy



## CNN Recipe

Recipe: **convolution** + subsampling (pooling) + hierarchy

### Definitional Note

If you have a background in signal processing or math, you may have already heard of convolution. However, the definitions in other domains and the one we use here are slightly different. The more common definition involves flipping the kernel horizontally and vertically before sliding.

**For our purposes, no flipping is needed. Flipping does not affect CNN's learning performance. If you are familiar with conventions involving flipping, just assume the kernel is pre-flipped.**

## CNN Recipe

Recipe: **convolution** + subsampling (pooling) + hierarchy

Let's go to section-1 in tutorial-1 to practice it!



## CNN Recipe

Recipe: ~~convolution~~ + **subsampling (pooling)** + hierarchy

Recall: filters give us global invariance

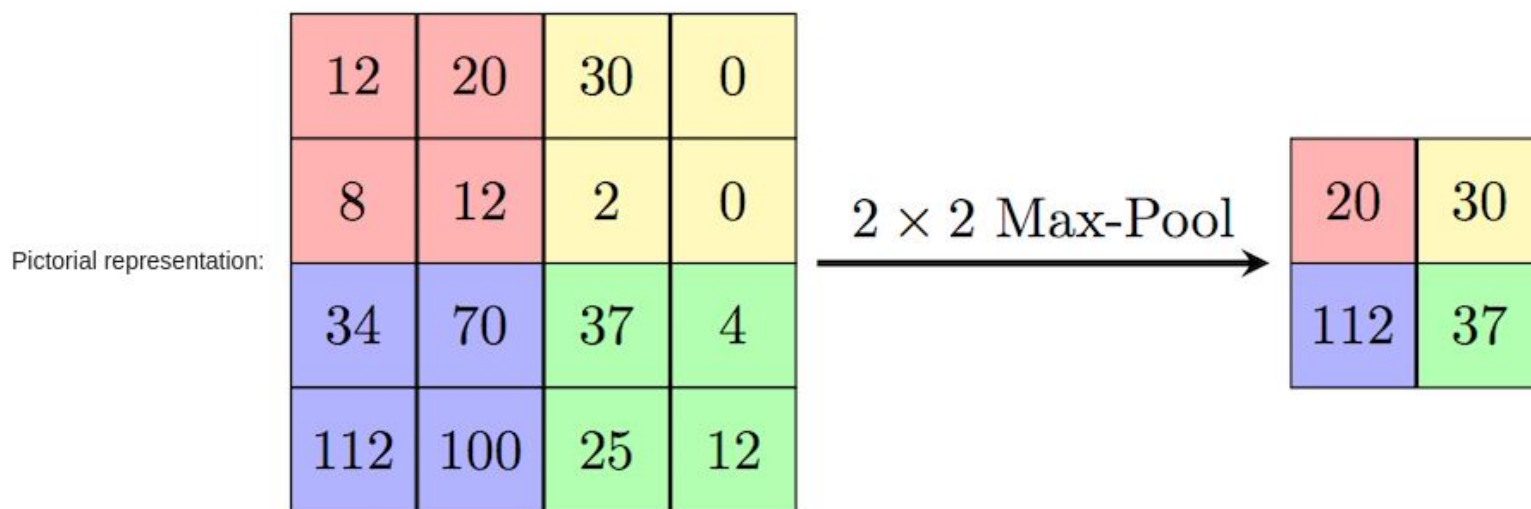
Pooling gives us local invariance



## CNN Recipe

Recipe: ~~convolution~~ + **subsampling (pooling)** + hierarchy

### Max-Pooling

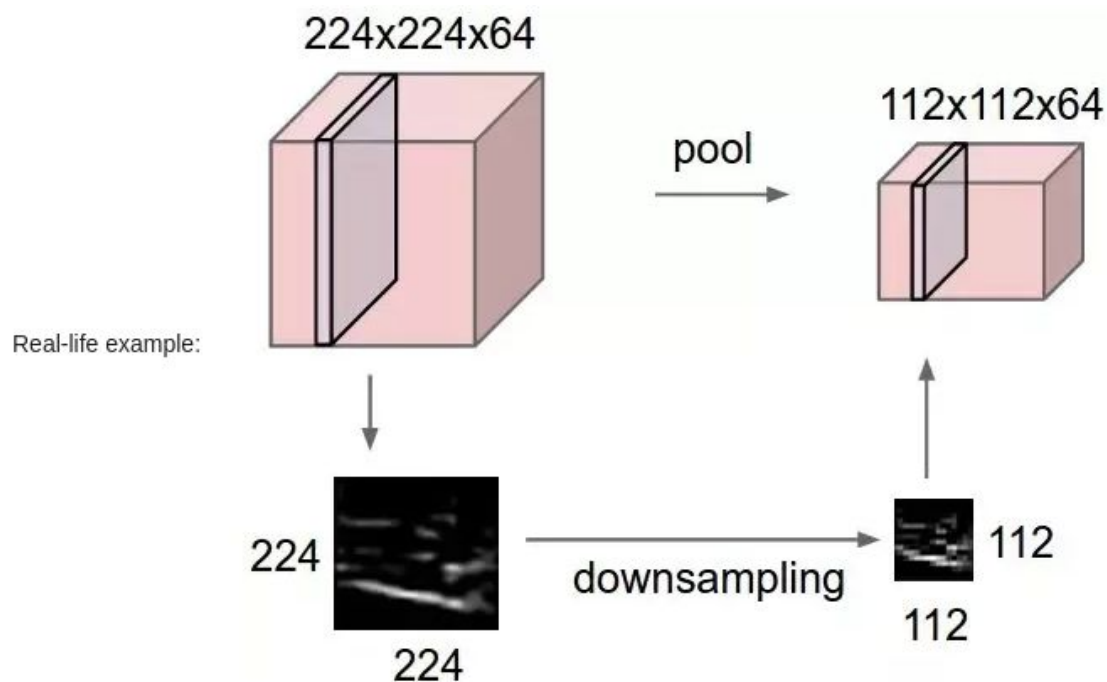




## CNN Recipe

Recipe: ~~convolution~~ + **subsampling (pooling)** + hierarchy

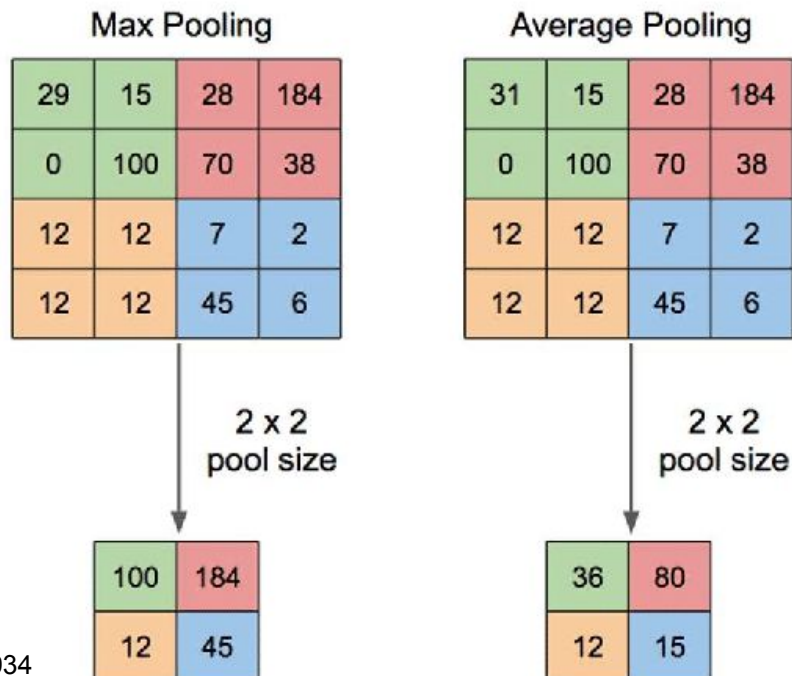
### Max-Pooling



## CNN Recipe

Recipe: ~~convolution~~ + **subsampling (pooling)** + hierarchy

### Average-Pooling

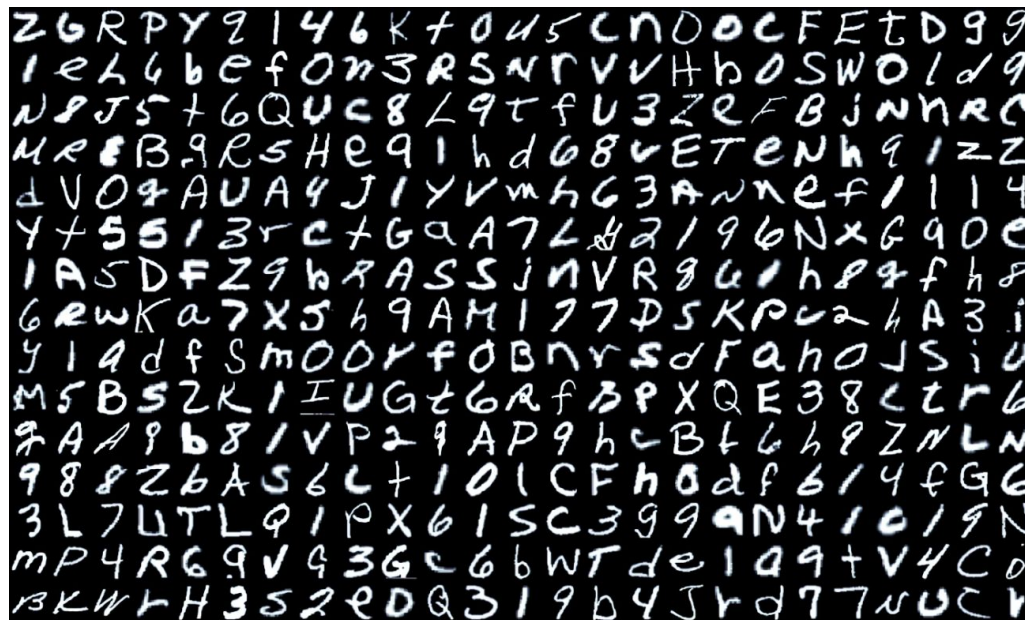


Source:  
<https://www.researchgate.net/publication/333593451/figure/download/fig2/AS:765890261966848@1559613876098/Illustration-of-Max-Pooling-and-Average-Pooling-Figure-2-above-shows-an-example-of-max.png>

## CNN Recipe

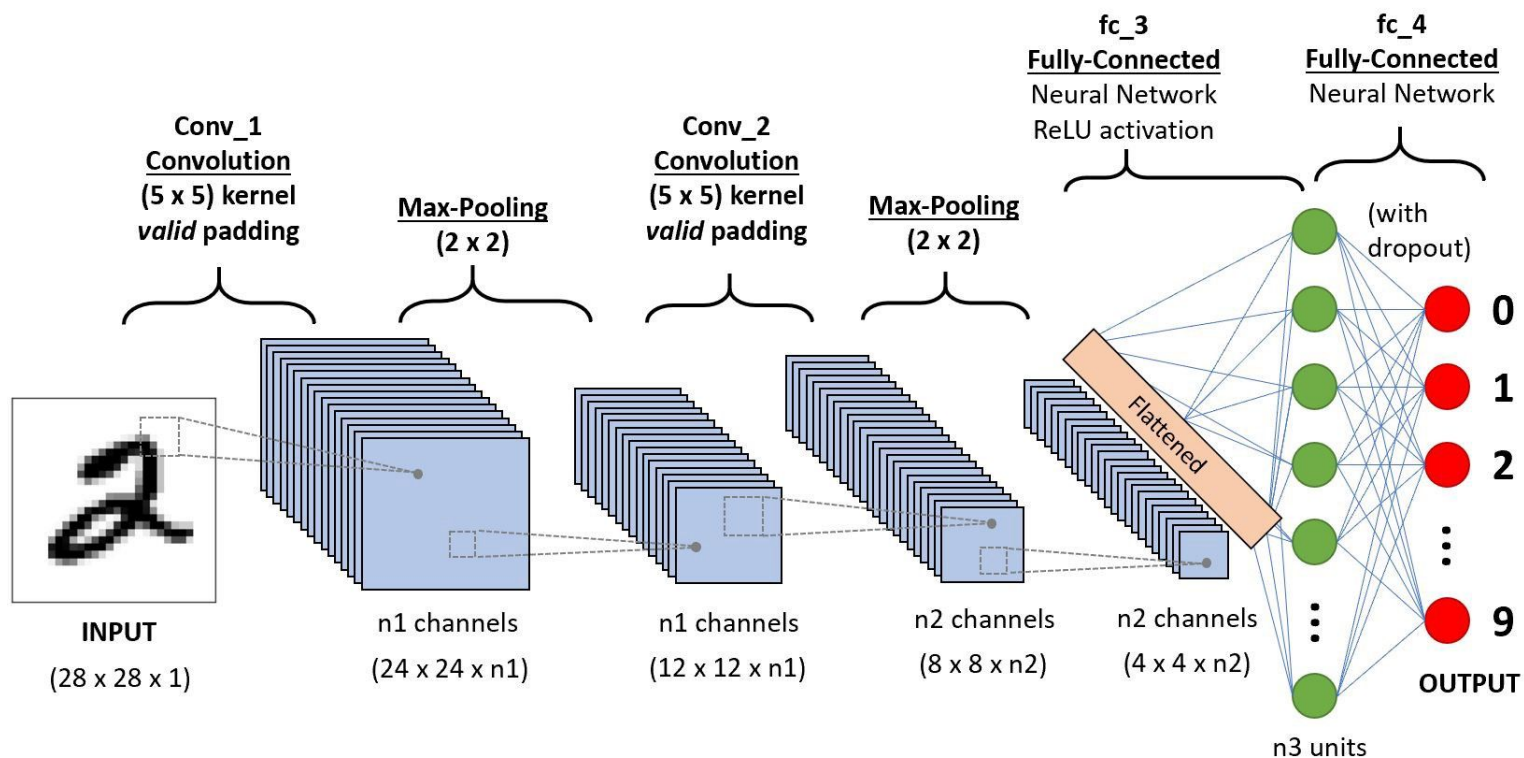
Recipe: ~~convolution~~ + **subsampling (pooling)** + hierarchy

Let's go to section-2 in tutorial-1 to practice it!



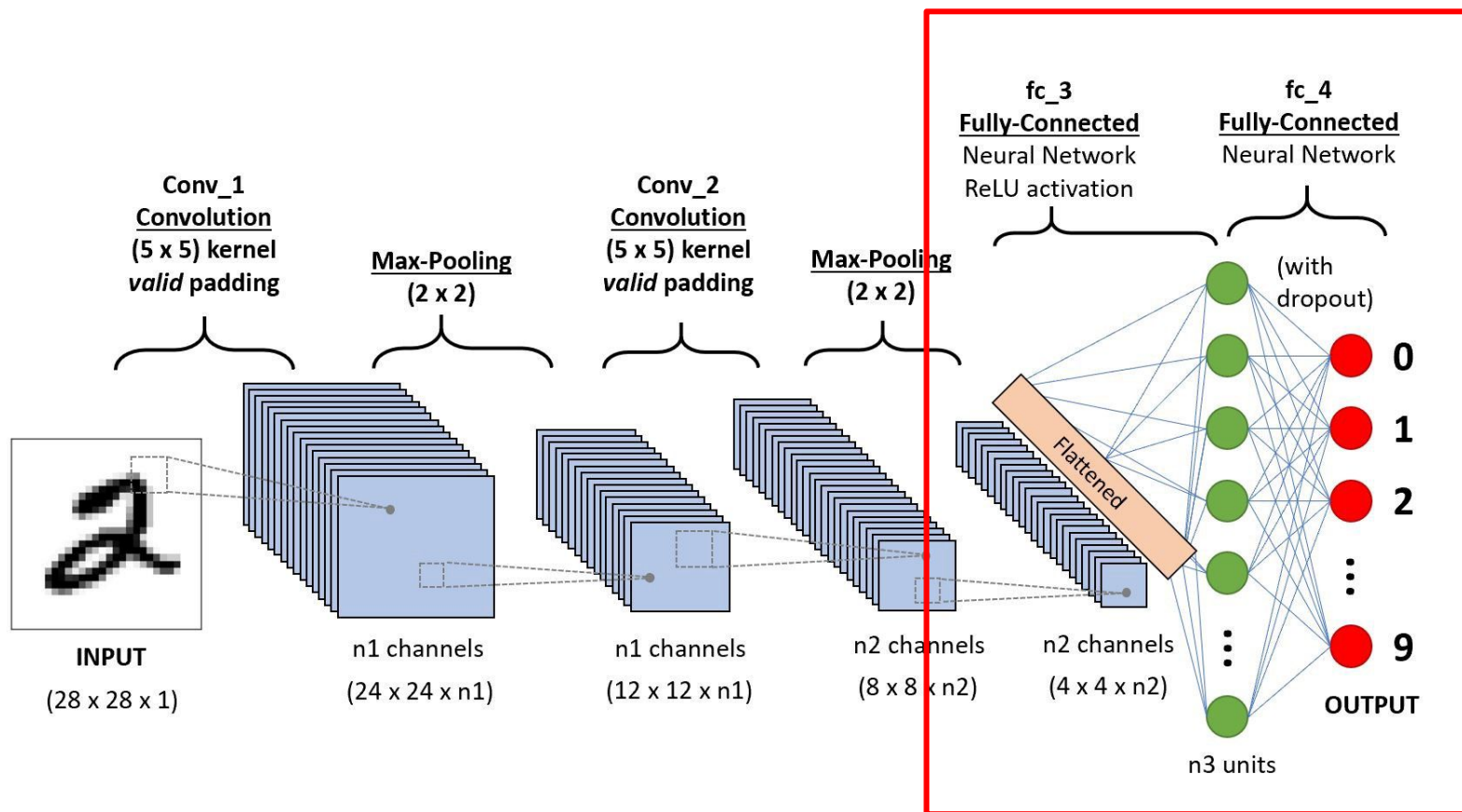
## CNN Recipe

Recipe: ~~convolution~~ + ~~subsampling (pooling)~~ + **hierarchy**



## CNN Recipe

Recipe: ~~convolution~~ + ~~subsampling (pooling)~~ + **hierarchy**



## CNN Recipe

Recipe: ~~convolution~~ + ~~subsampling (pooling)~~ + **hierarchy**

Let's put everything together in Section 3 of tutorial 1!!

