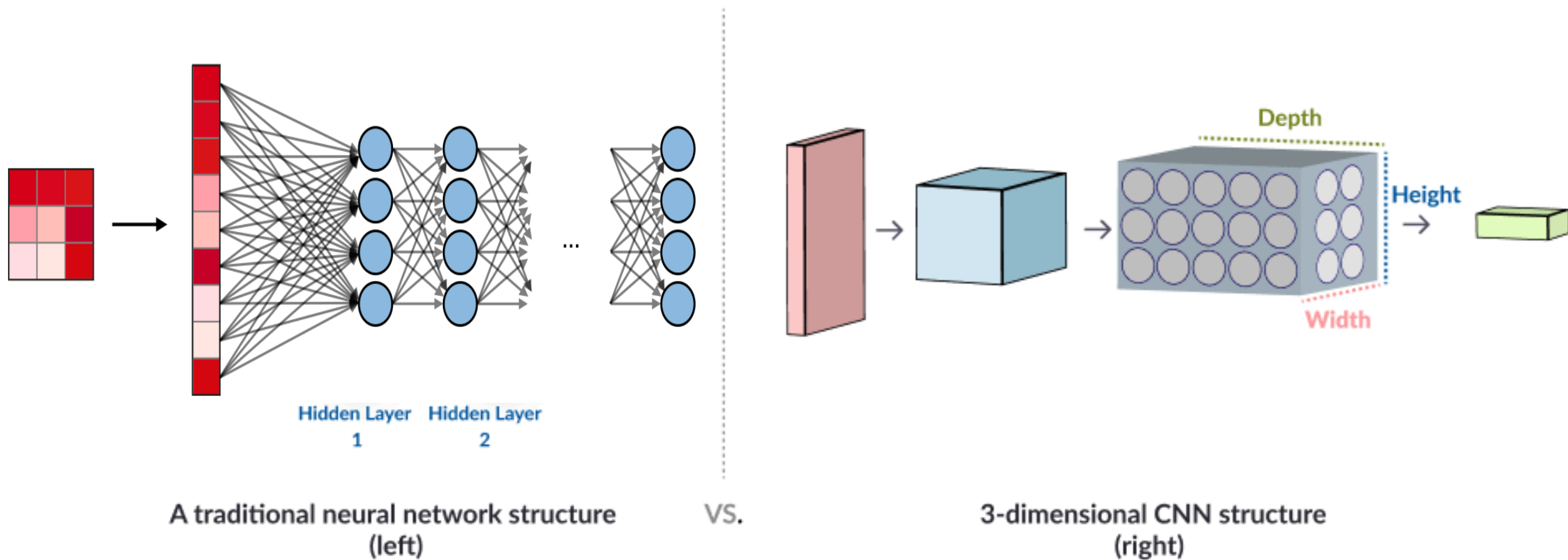


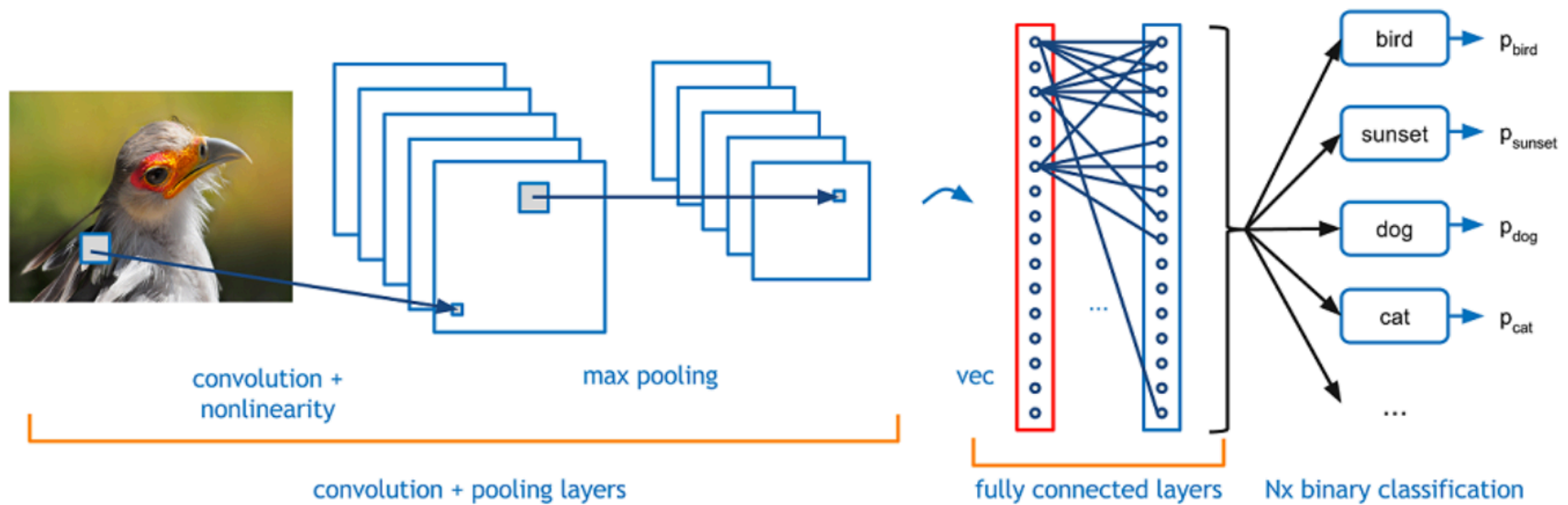
Convolution Neural Network

- Pramod Divakarmurthy

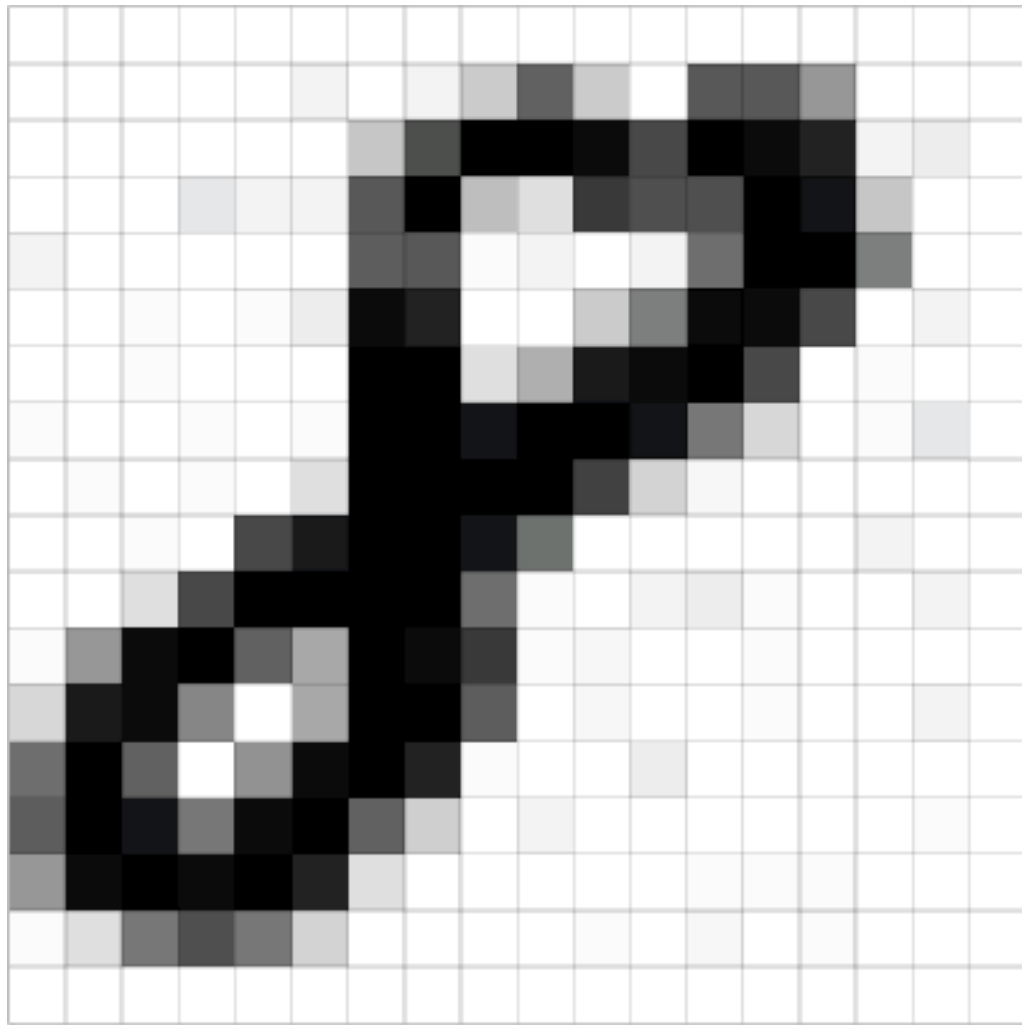
Traditional Neural Network Vs Convolution Neural Network



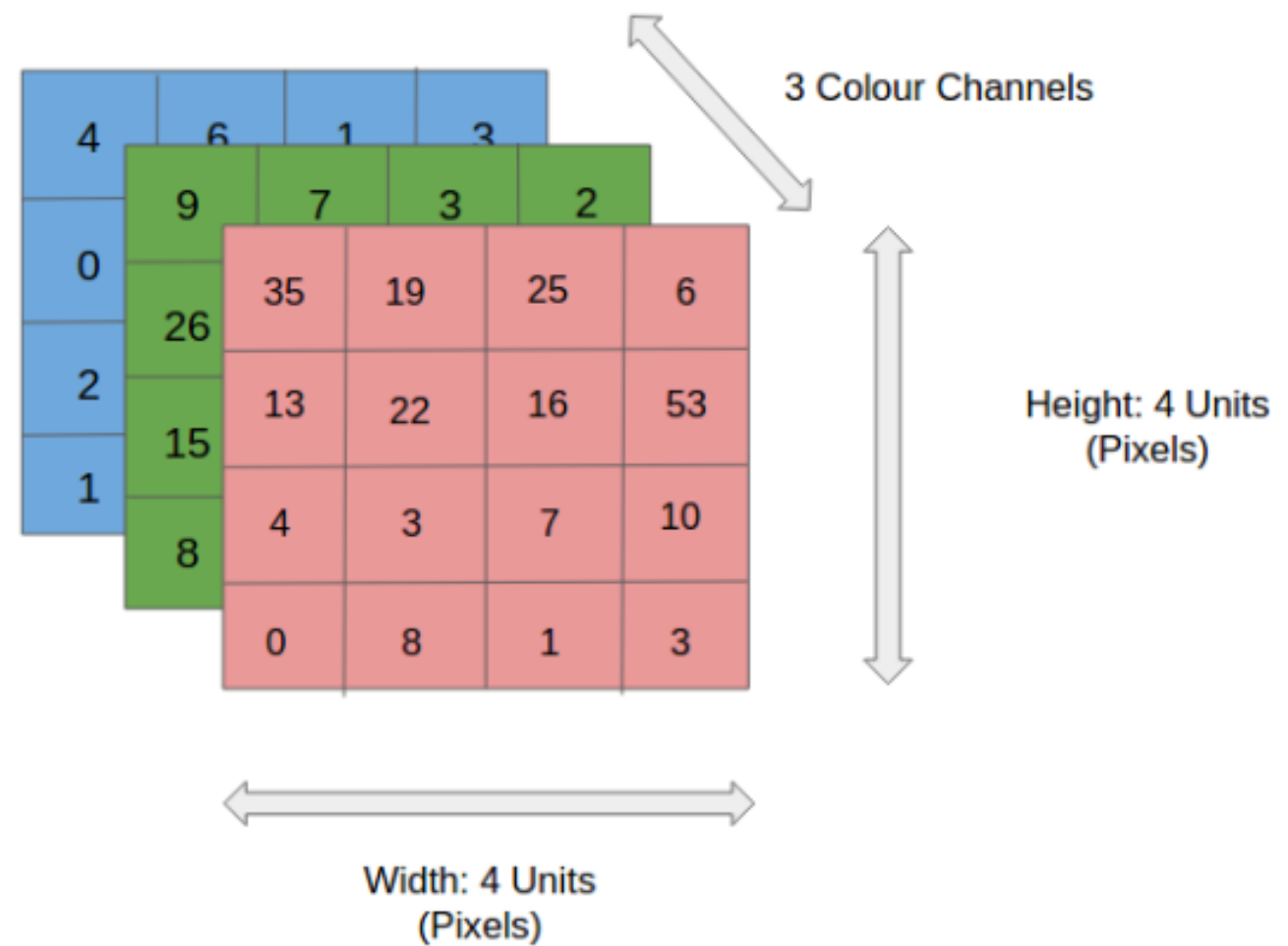
CNN And Layers



1. Input Layer - An Image is a matrix of pixel values

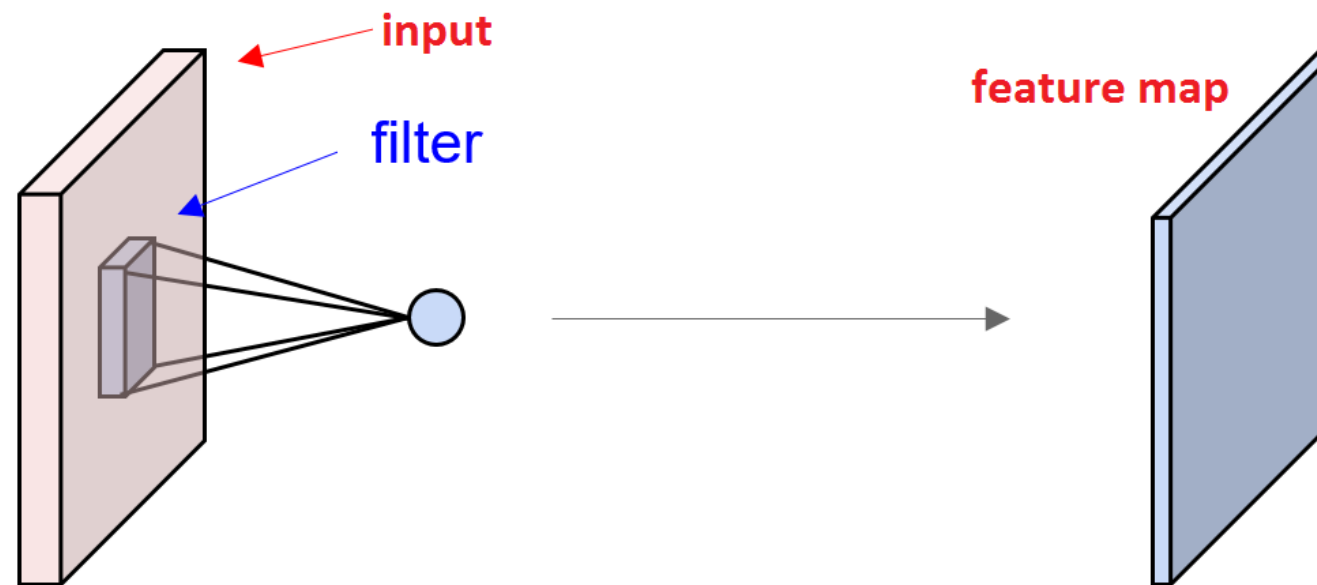


GrayScale



RGB

2. Convolution Layer — Convolution



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

Image

*

1	0	1
0	1	0
1	0	1

Kernel/Filter (K)

=

1 _{x1}	1 _{x0}	1 _{x1}	0	0
0 _{x0}	1 _{x1}	1 _{x0}	1	0
0 _{x1}	0 _{x0}	1 _{x1}	1	1
0	0	1	1	0
0	1	1	0	0

Image

4		

Convolved
Feature

Convolution Operation on a MxNx3 image matrix with a 3x3x3 Kernel

0	0	0	0	0	0	...
0	156	155	156	158	158	...
0	153	154	157	159	159	...
0	149	151	155	158	159	...
0	146	146	149	153	158	...
0	145	143	143	148	158	...
...

Input Channel #1 (Red)

0	0	0	0	0	0	...
0	167	166	167	169	169	...
0	164	165	168	170	170	...
0	160	162	166	169	170	...
0	156	156	159	163	168	...
0	155	153	153	158	168	...
...

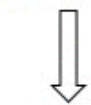
Input Channel #2 (Green)

0	0	0	0	0	0	...
0	163	162	163	165	165	...
0	160	161	164	166	166	...
0	156	158	162	165	166	...
0	155	155	158	162	167	...
0	154	152	152	157	167	...
...

Input Channel #3 (Blue)

-1	-1	1
0	1	-1
0	1	1

Kernel Channel #1



308

1	0	0
1	-1	-1
1	0	-1

Kernel Channel #2



-498

0	1	1
0	1	0
1	-1	1

Kernel Channel #3



164

+

+



Bias = 1

+ 1 = -25

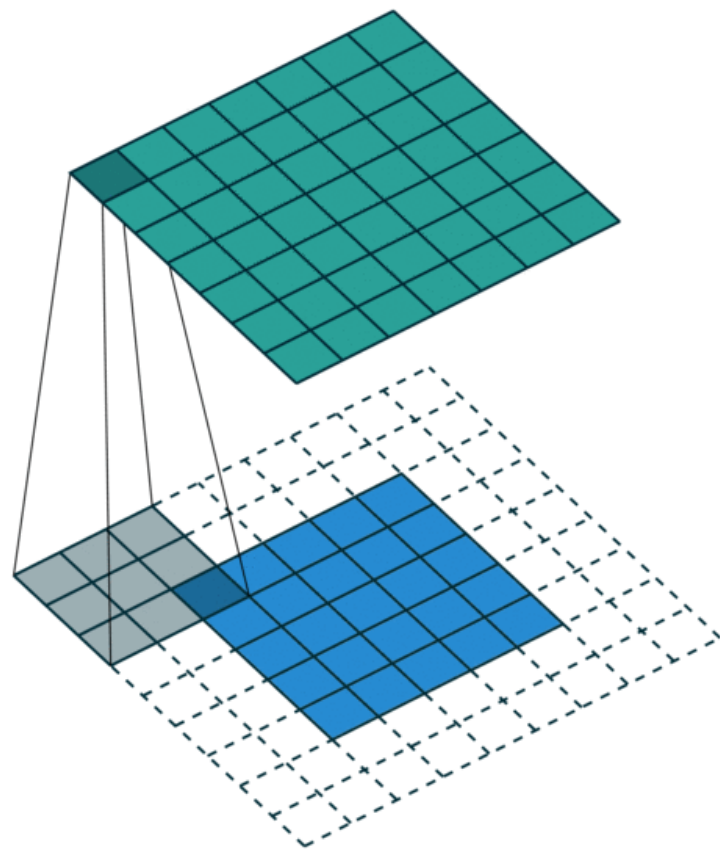
Output

-25				...
				...
				...
				...
...

Padding

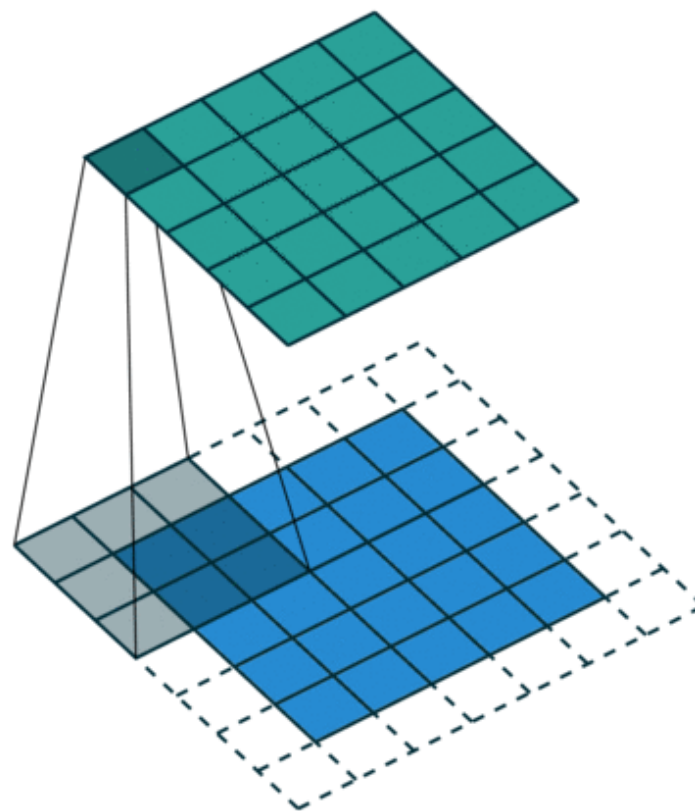
- Pixels in the middle are used more often than pixels on corners and edges
- Information on the borders of images are not preserved

Full Padding



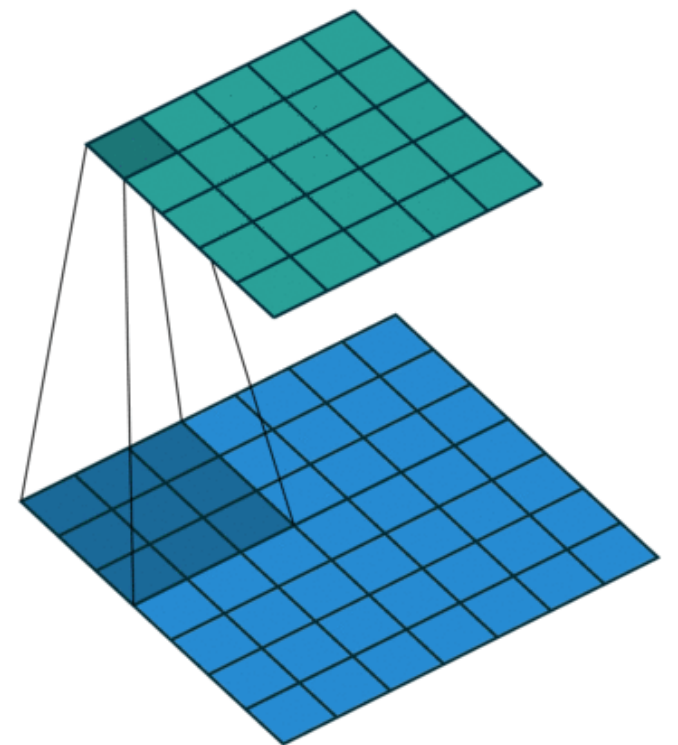
Output size
==
 $(m+k-1) * (m+k-1)$

Same Padding



Output size
==
Image size

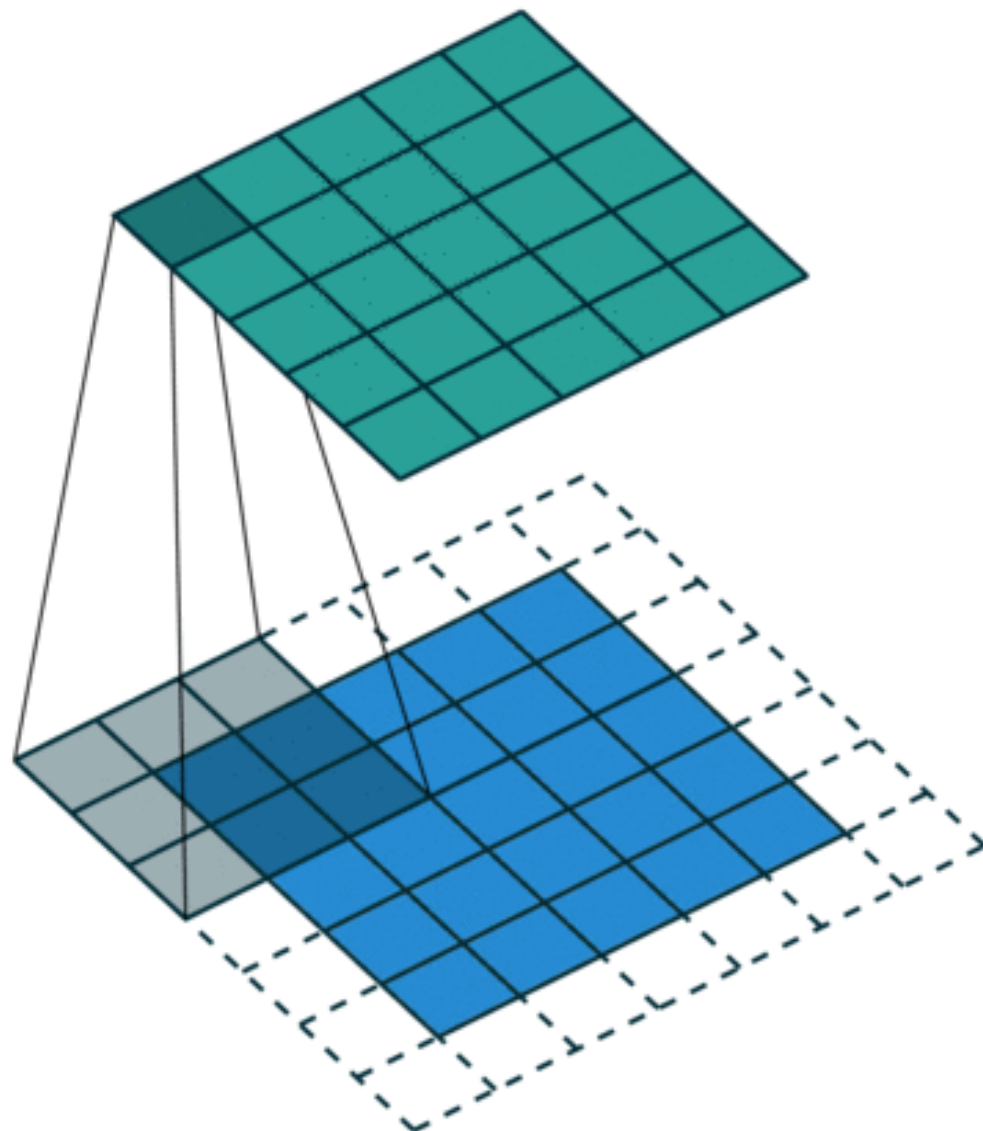
Valid Padding



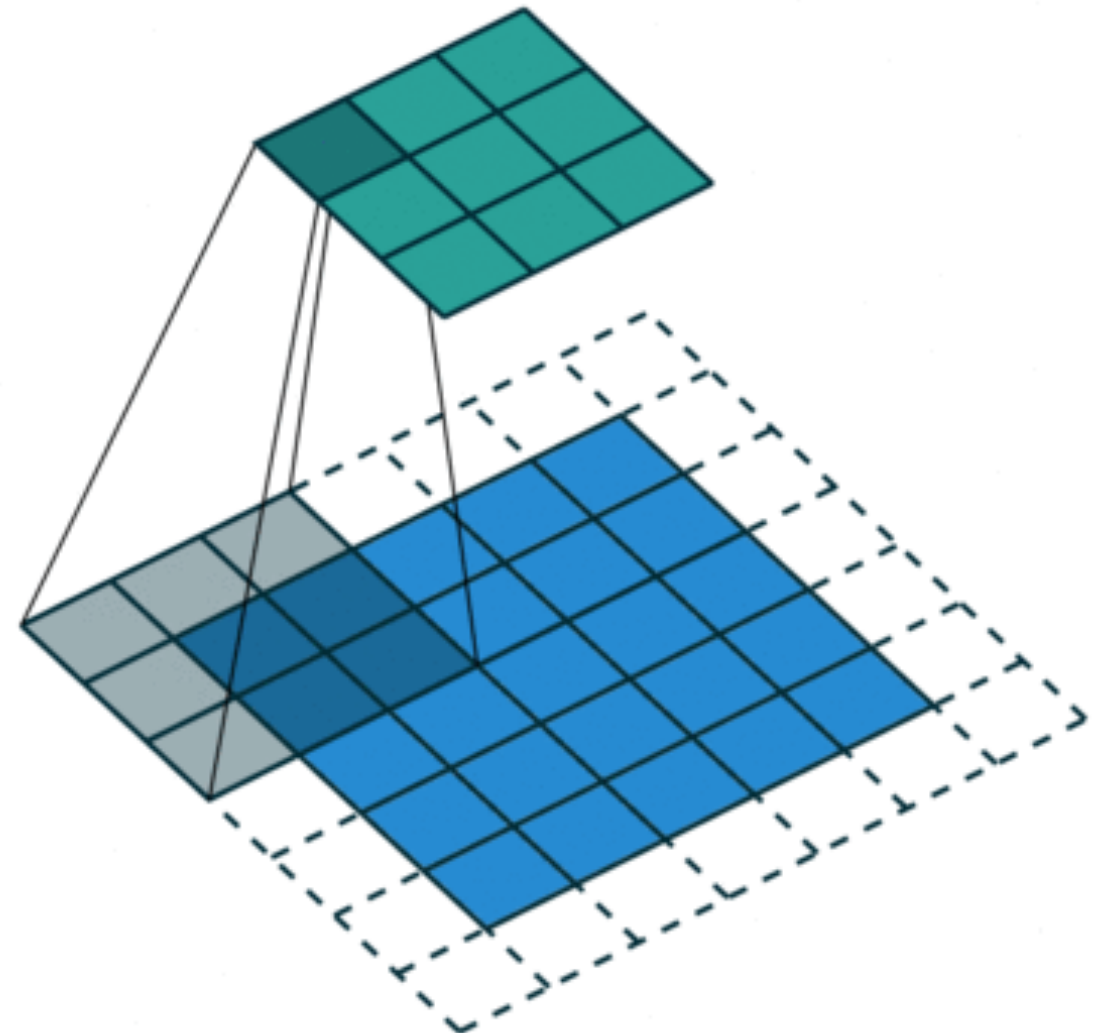
Output size
==
 $(m-k+1) * (m-k+1)$

Stride

- The number of pixels by which we slide our filter matrix over the input matrix
- Having a larger stride will produce smaller feature maps.



Stride Length = 1 (Default)



Stride Length = 2

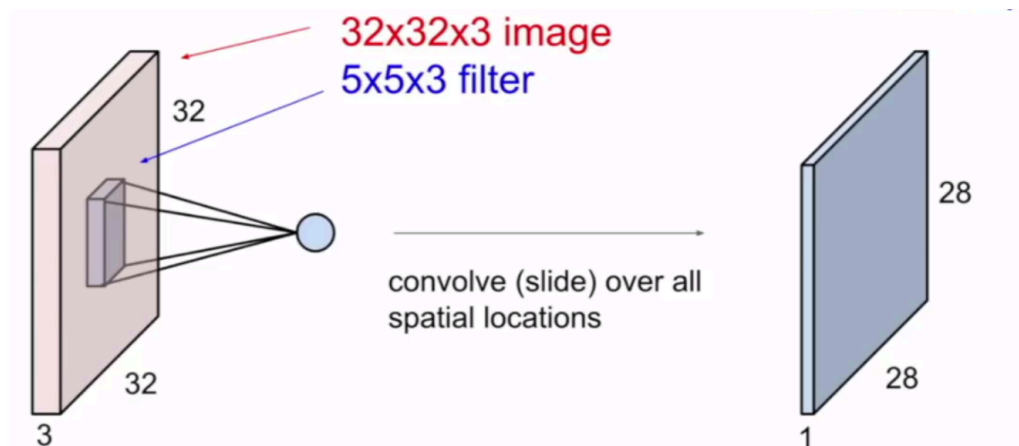
Size of Activation Map

There is a formula which is used in determining the dimension of the activation maps:

$$(N + 2P - F) / S + 1$$

where

- N = Dimension of image (input) file
- P = Padding
- F = Dimension of filter
- S = Stride

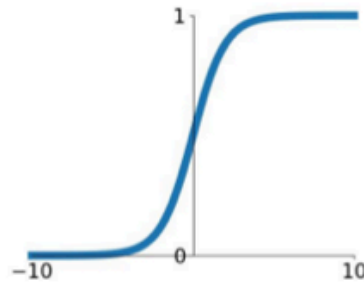


$$\text{Activation Map Size} = (32 + 0 - 5) / 1 + 1 = 28$$

Activation Function

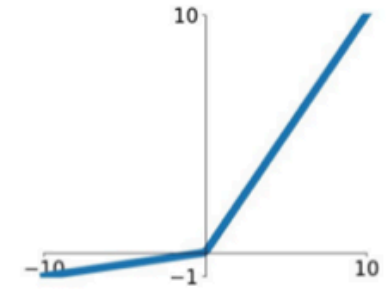
Sigmoid

$$\sigma(x) = \frac{1}{1+e^{-x}}$$



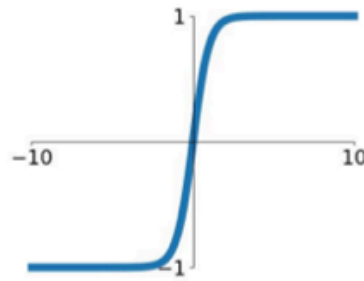
Leaky ReLU

$$\max(0.1x, x)$$



tanh

$$\tanh(x)$$

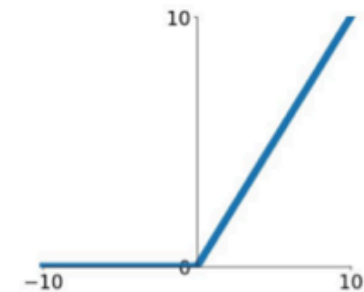


Maxout

$$\max(w_1^T x + b_1, w_2^T x + b_2)$$

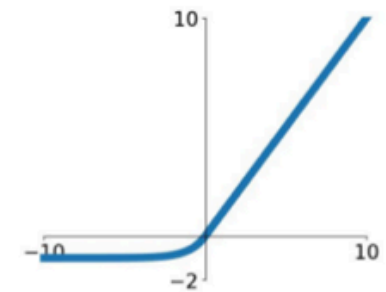
ReLU

$$\max(0, x)$$



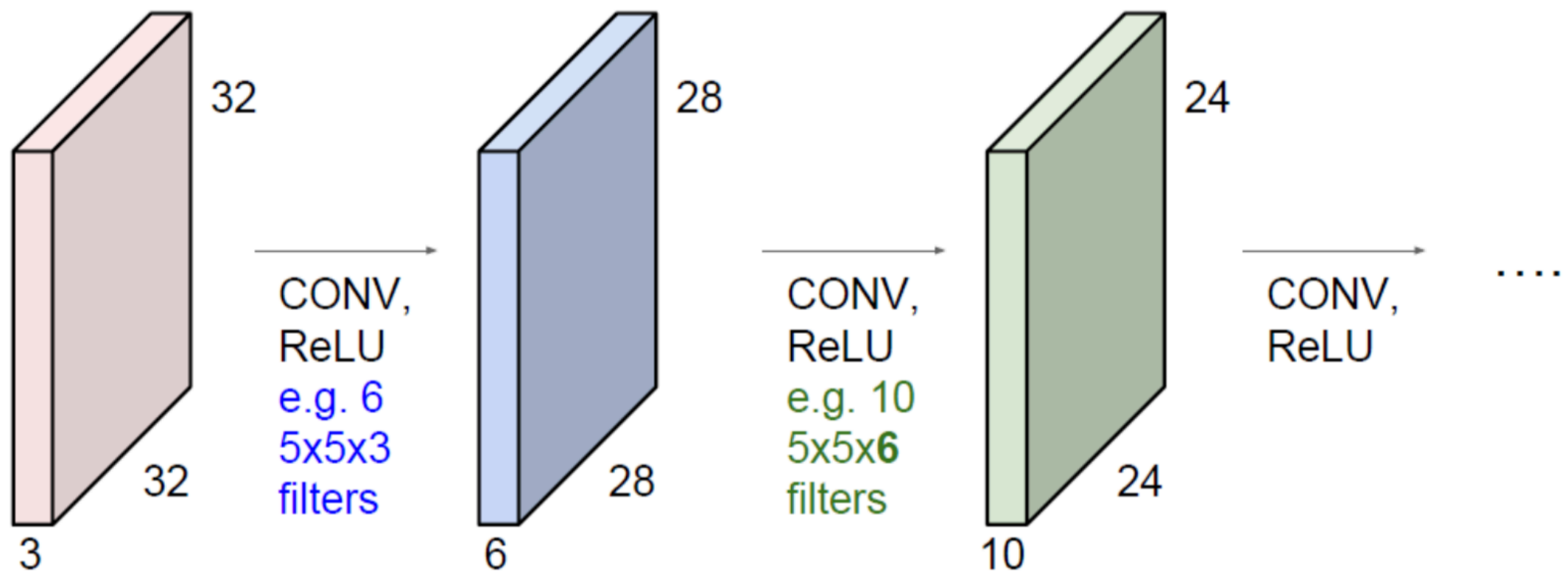
ELU

$$\begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$$



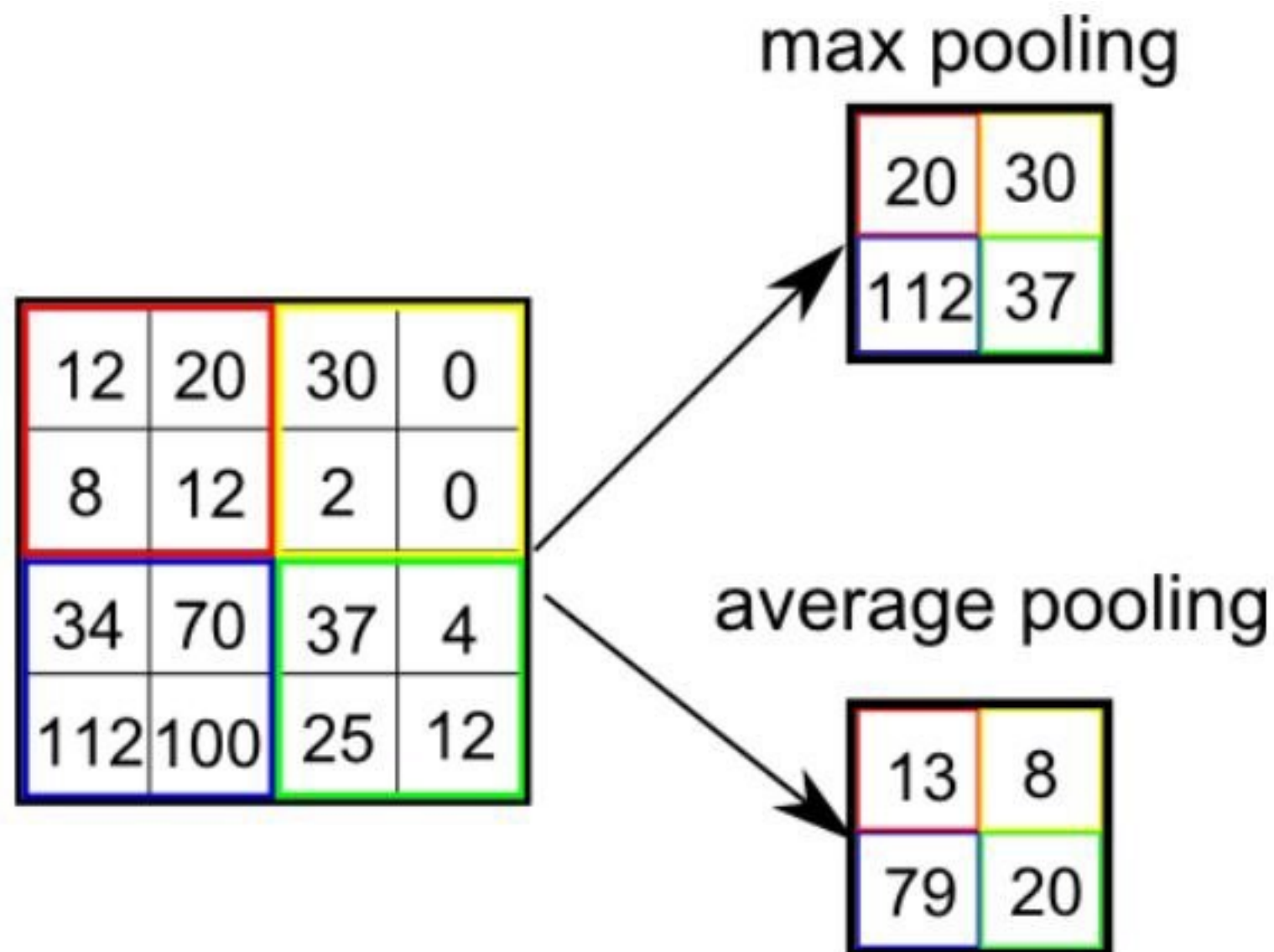
Stacking Convolution Layers

ConvNet is a sequence of convolutional layer, interspersed with activation function

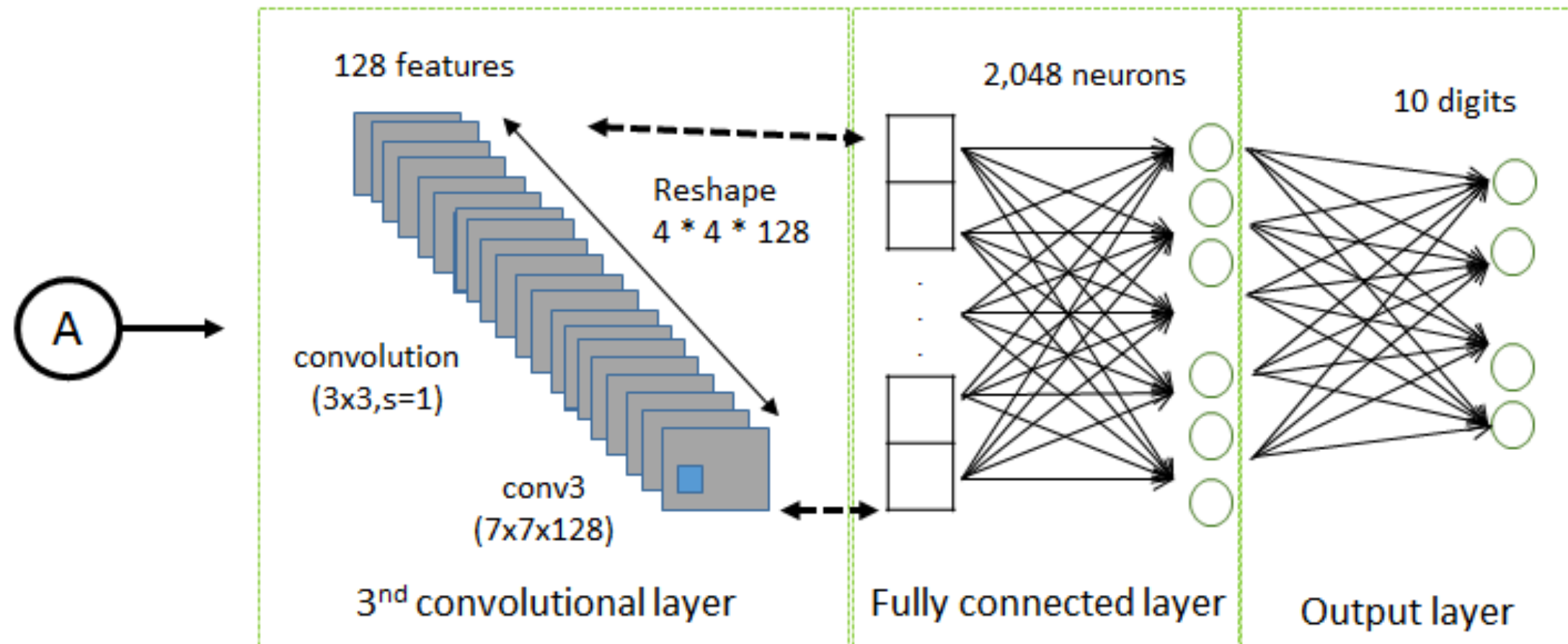


3. Pooling Layer

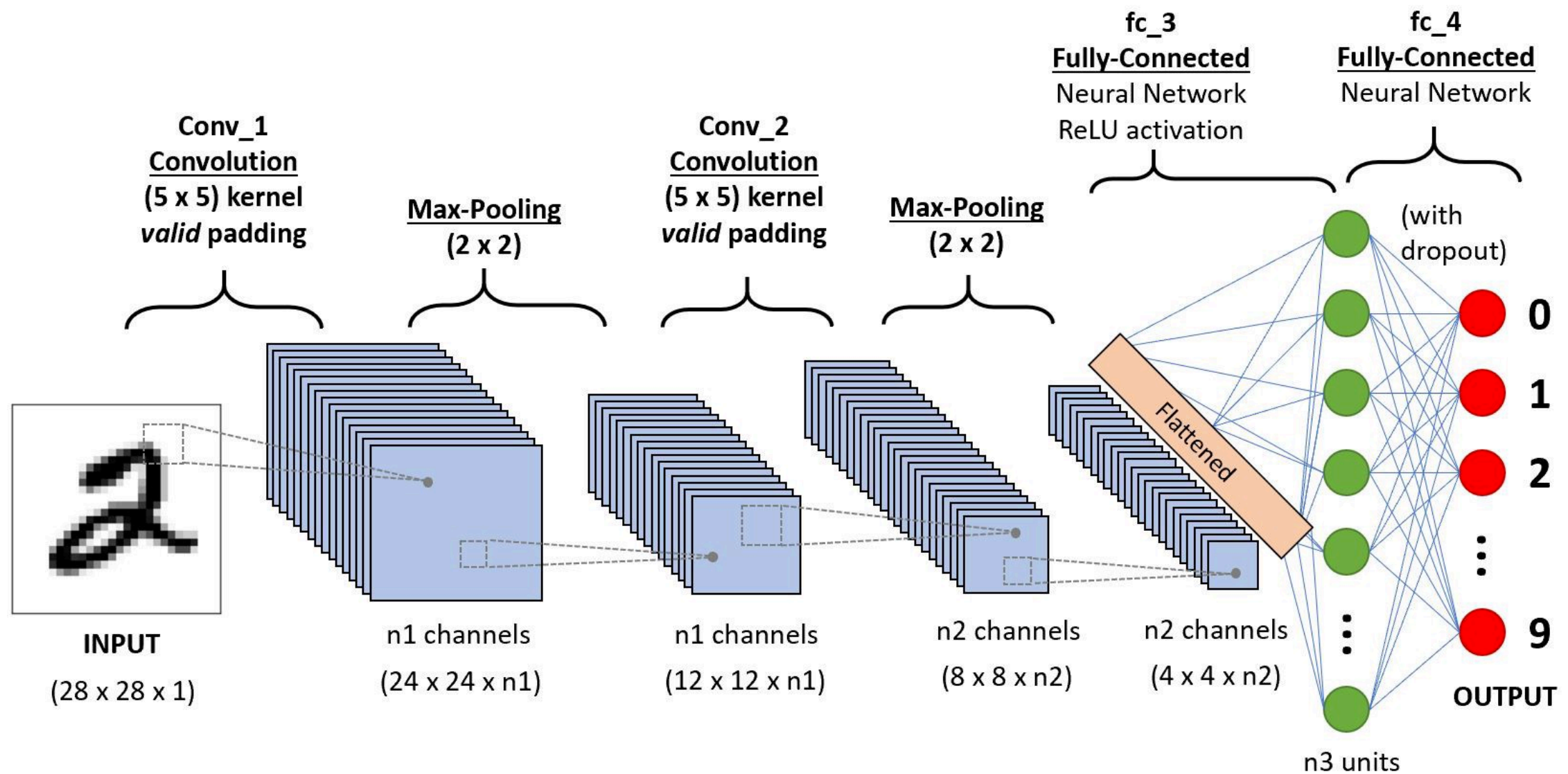
- To progressively reduce the spatial size (downsampling) of the Convolved Feature
- Reduces the number of parameters and computational power
- **Extracting dominant features** which are rotational and positional invariant



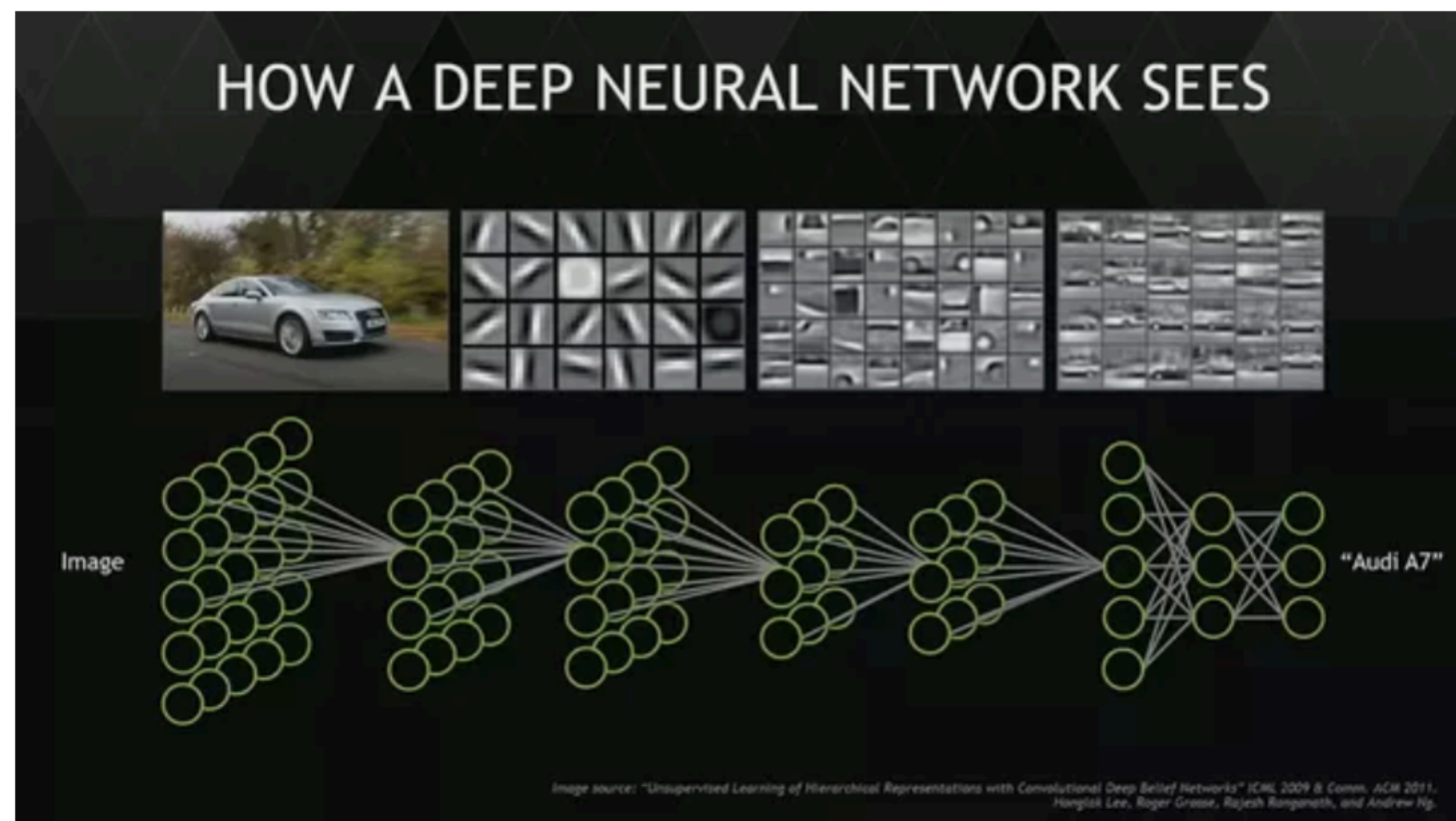
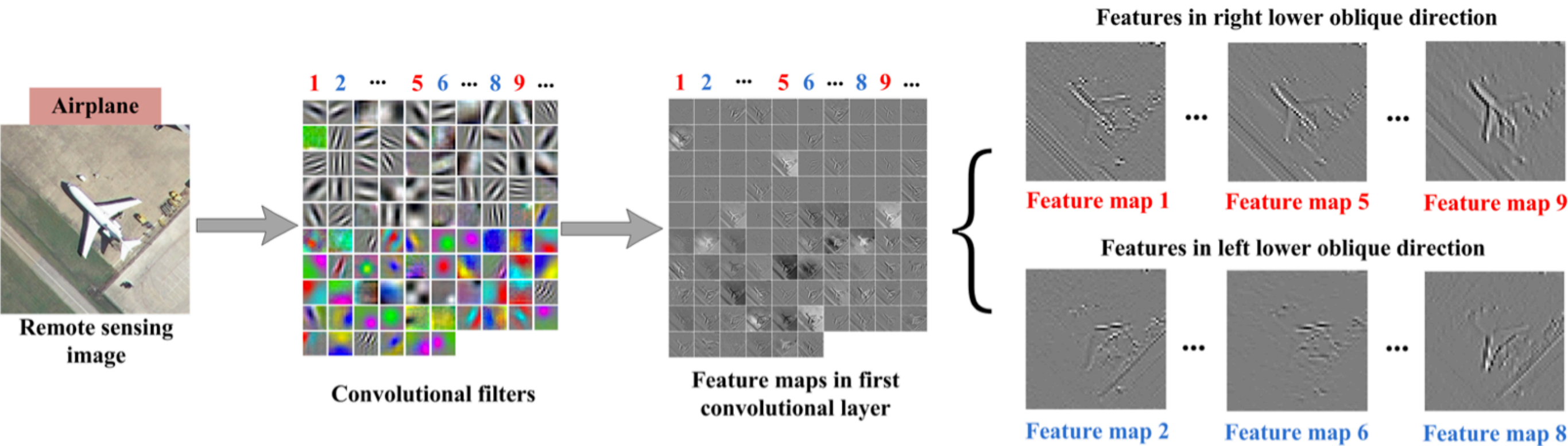
4. Fully Connected Layer (FC Layer)



CNN Sequence To Classify Handwritten Digits



Visualizing Filters & Feature Maps



Popular Architectures

Year	CNN	Developed by	Place	Top-5 error rate	No. of parameters
1998	LeNet(8)	Yann LeCun et al			60 thousand
2012	AlexNet(7)	Alex Krizhevsky, Geoffrey Hinton, Ilya Sutskever	1st	15.3%	60 million
2013	ZFNet()	Matthew Zeiler and Rob Fergus	1st	14.8%	
2014	GoogLeNet(19)	Google	1st	6.67%	4 million
2014	VGG Net(16)	Simonyan, Zisserman	2nd	7.3%	138 million
2015	<u>ResNet(152)</u>	Kaiming He	1st	3.6%	