



Foundations of Machine Learning (CS 725)

FALL 2024

Lecture 16:

- Convolutional Neural Networks

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Convolutional Neural Networks (CNNs)

- Fully connected (dense) layers have no awareness of spatial information
- Key concept behind convolutional layers is that of ***kernels*** or ***filters***
- Filters slide across an input space to detect spatial patterns (translation invariance) in local regions (locality)

Convolution Layer

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 Layer

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

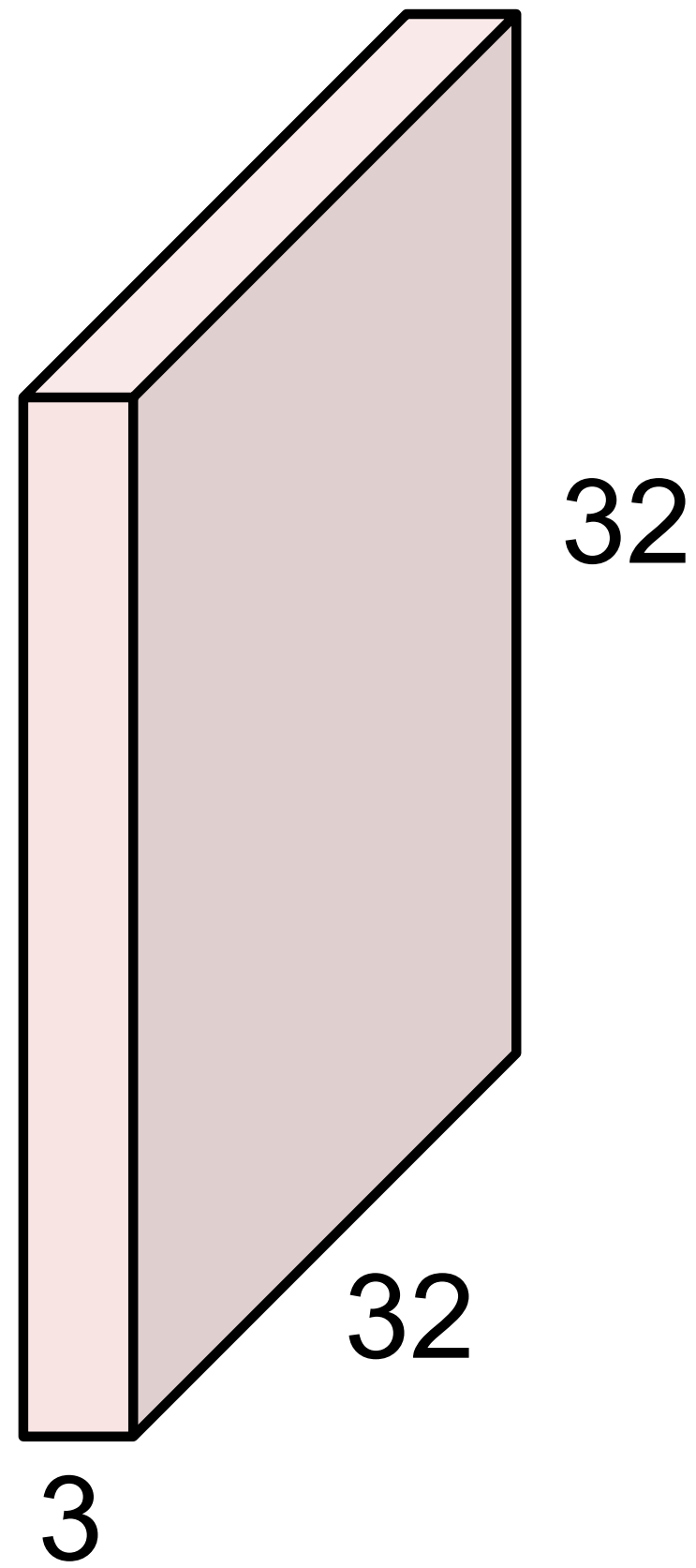
Image

4	3	4
2	4	3
2	3	4

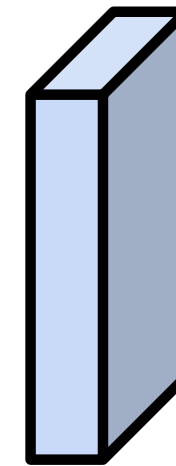
Convolved
Feature

Convolution Layer

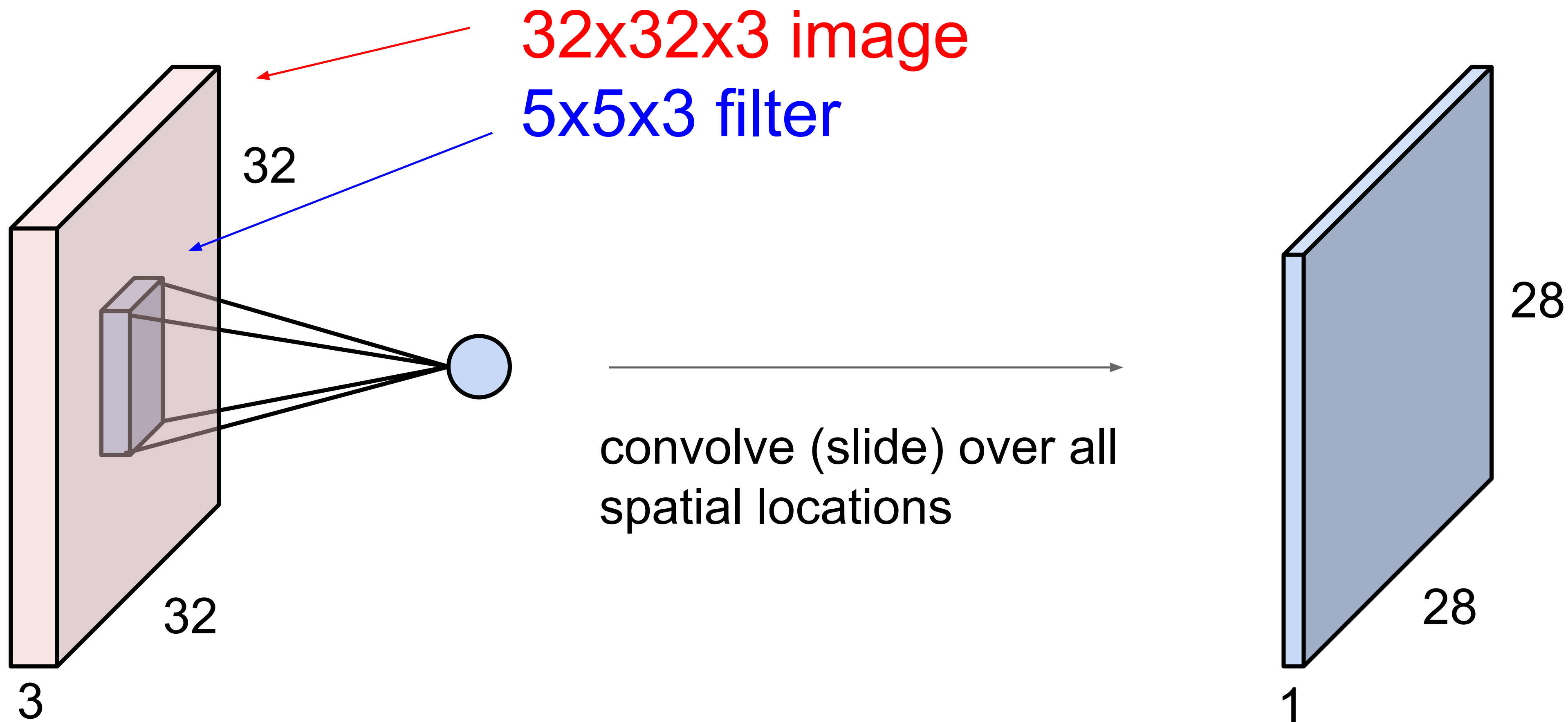
32x32x3 image



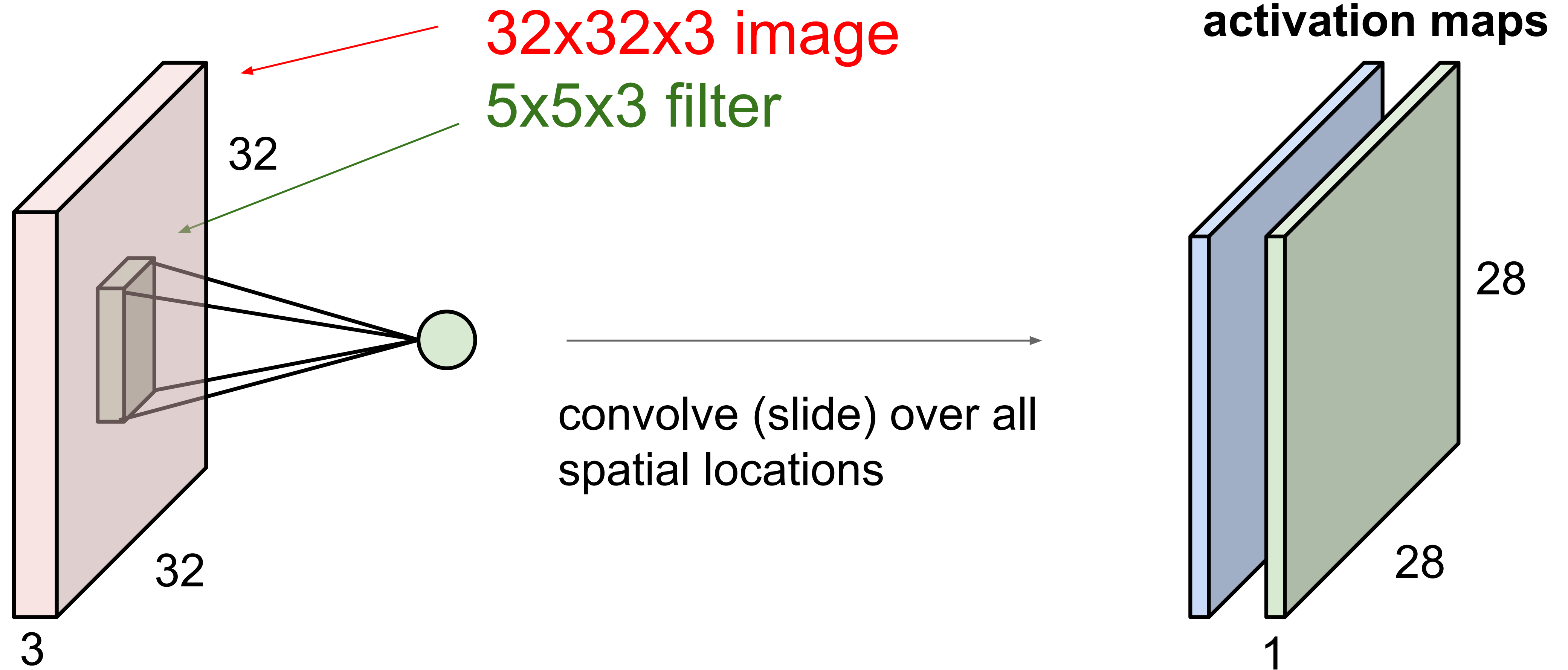
5x5x3 filter



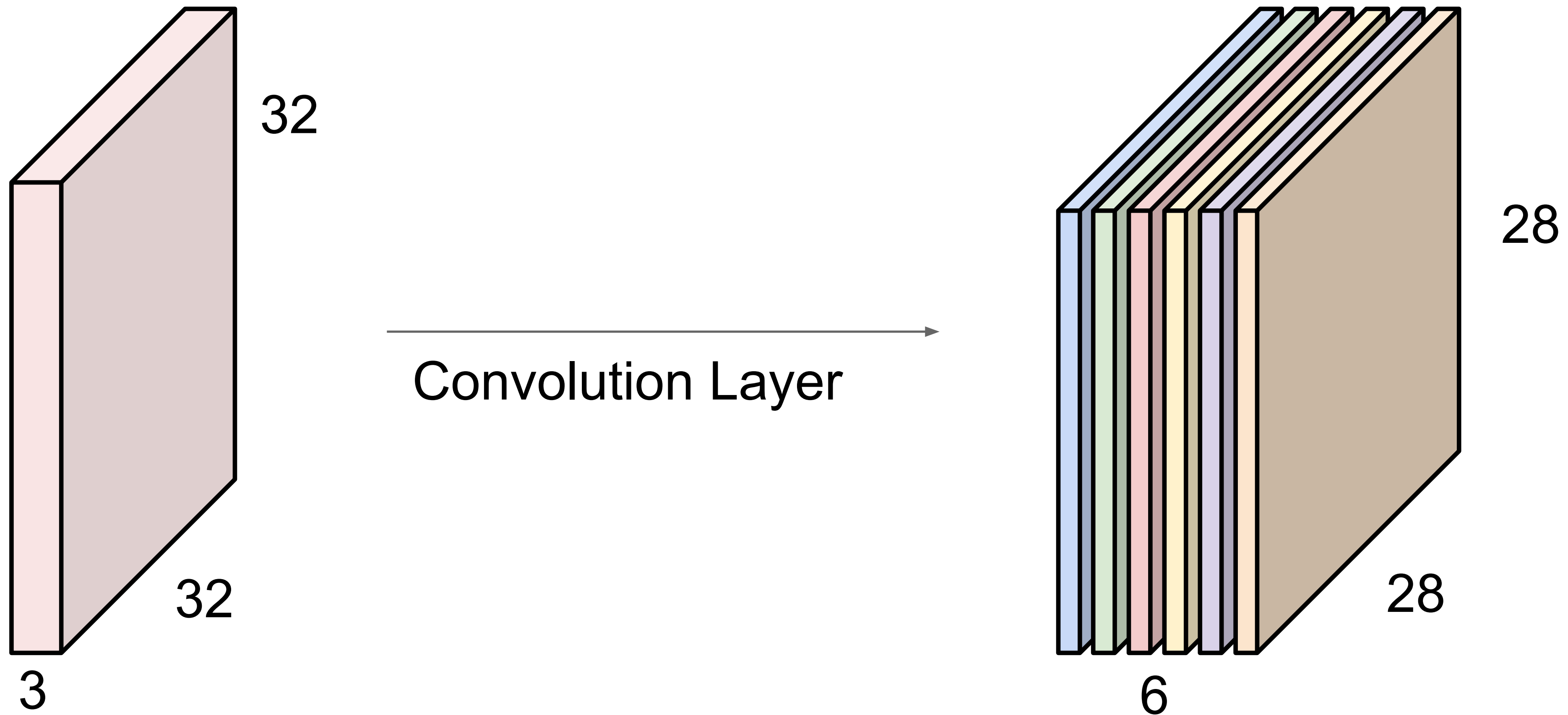
Convolution Layer



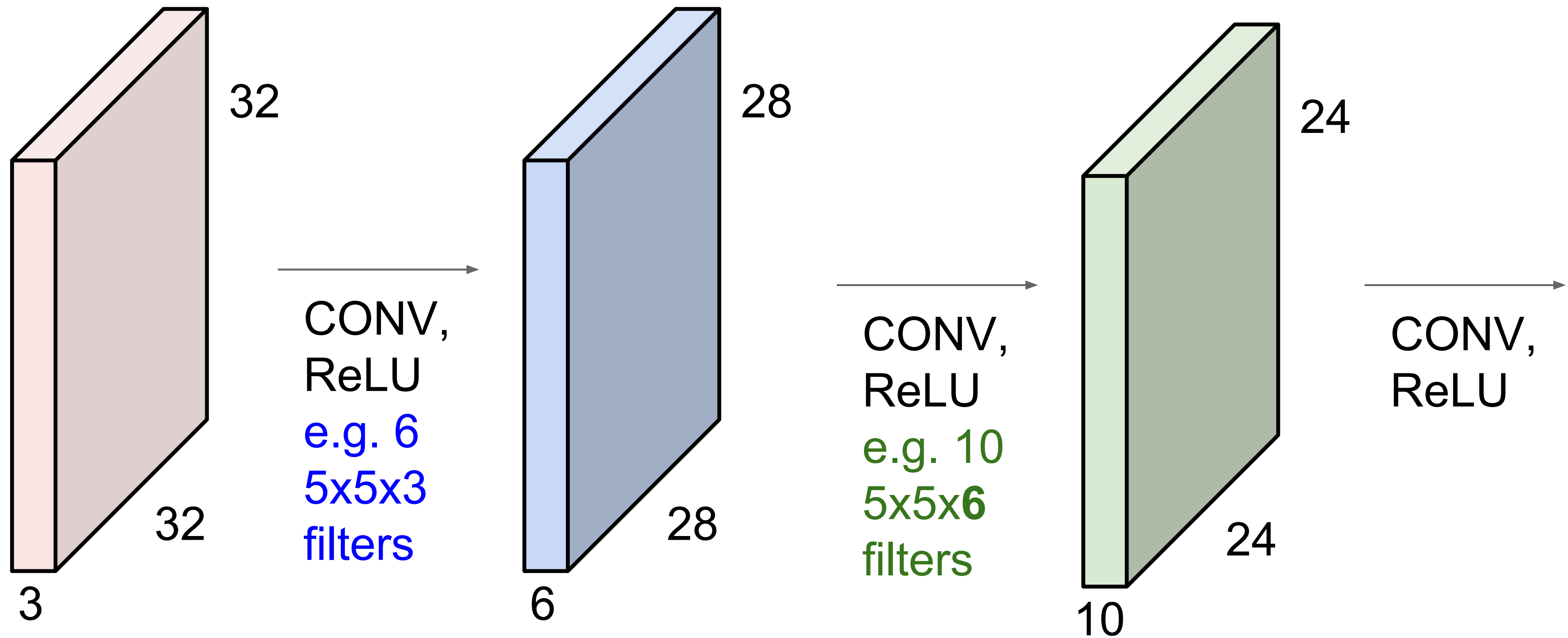
Convolution Layer



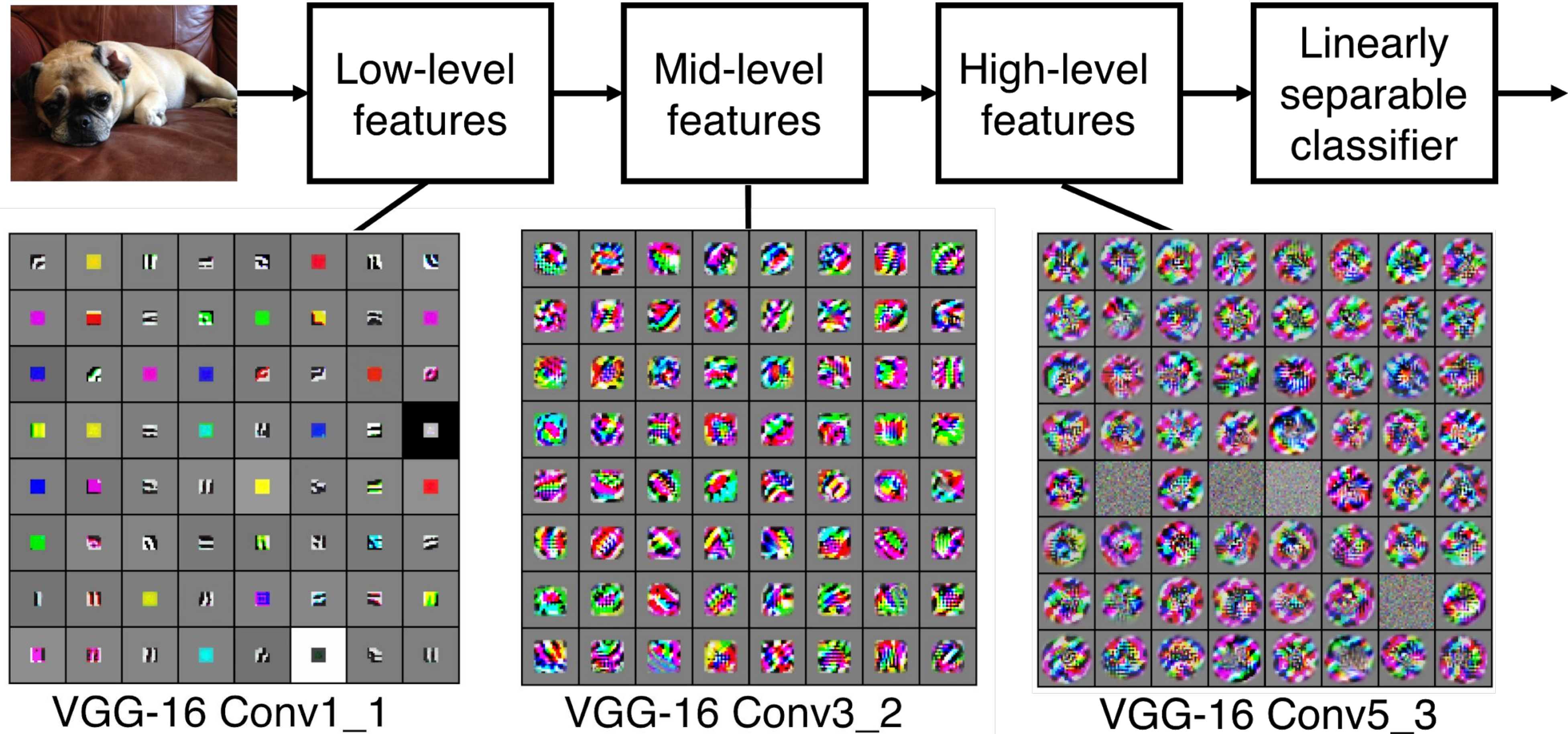
Convolution Layer



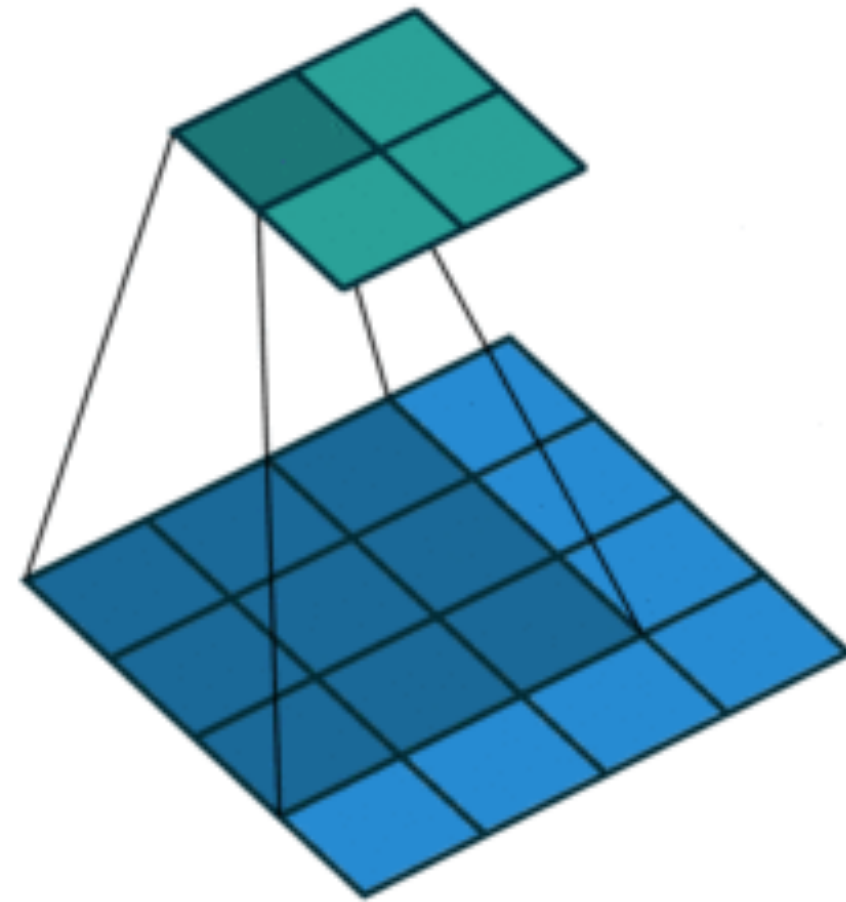
Convolutional Neural Network



What do these layers learn?

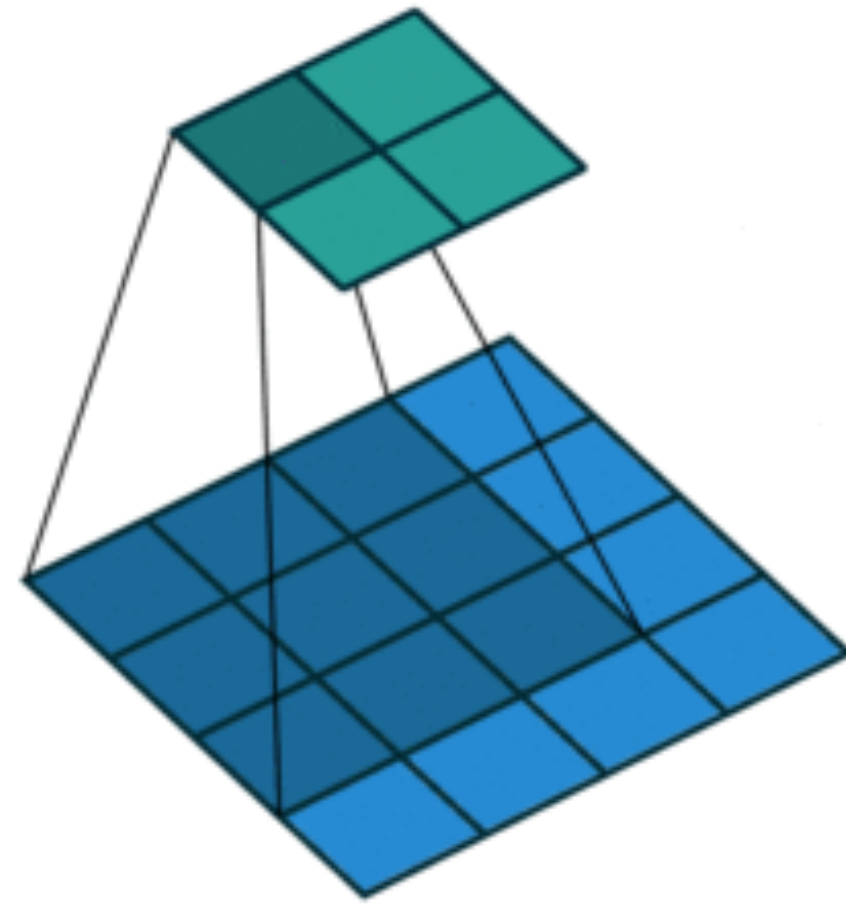


Convolutional Neural Networks (CNNs)

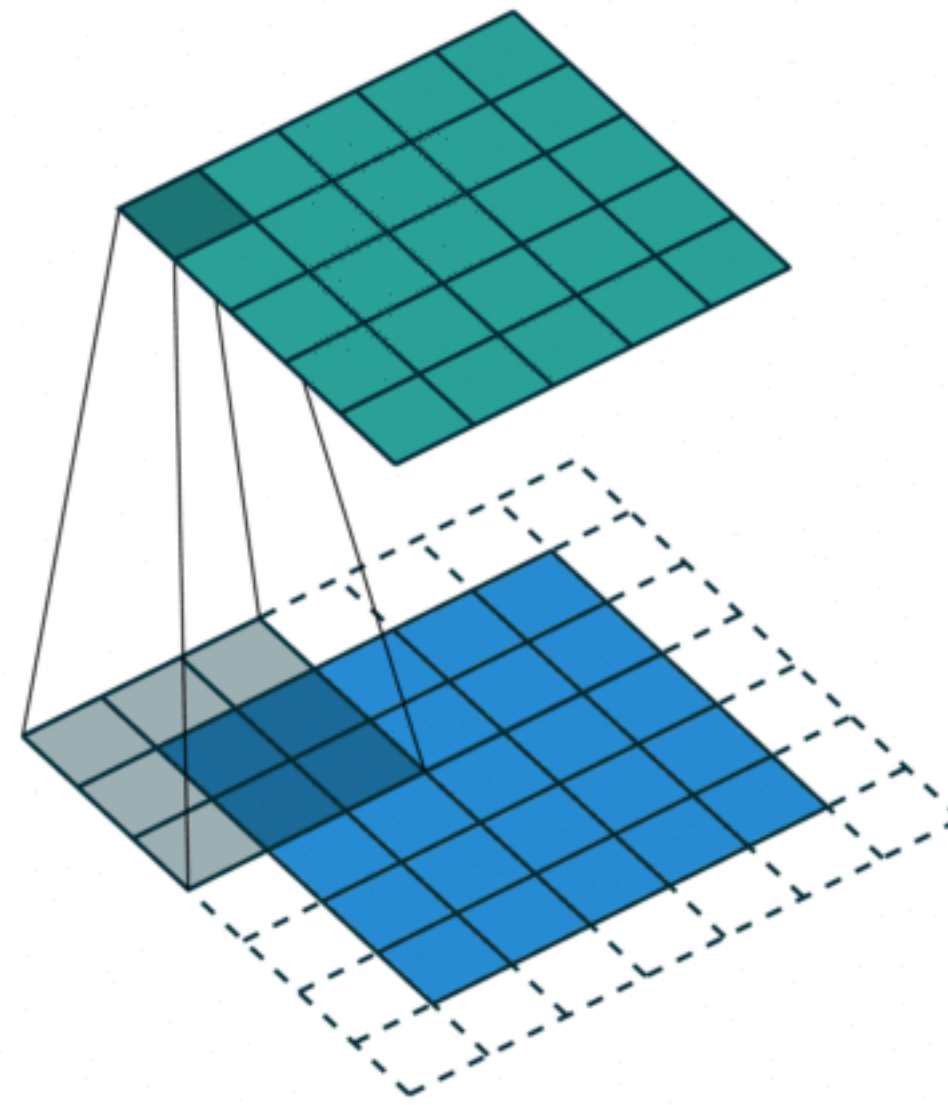


Stride=1, No padding

Convolutional Neural Networks (CNNs)

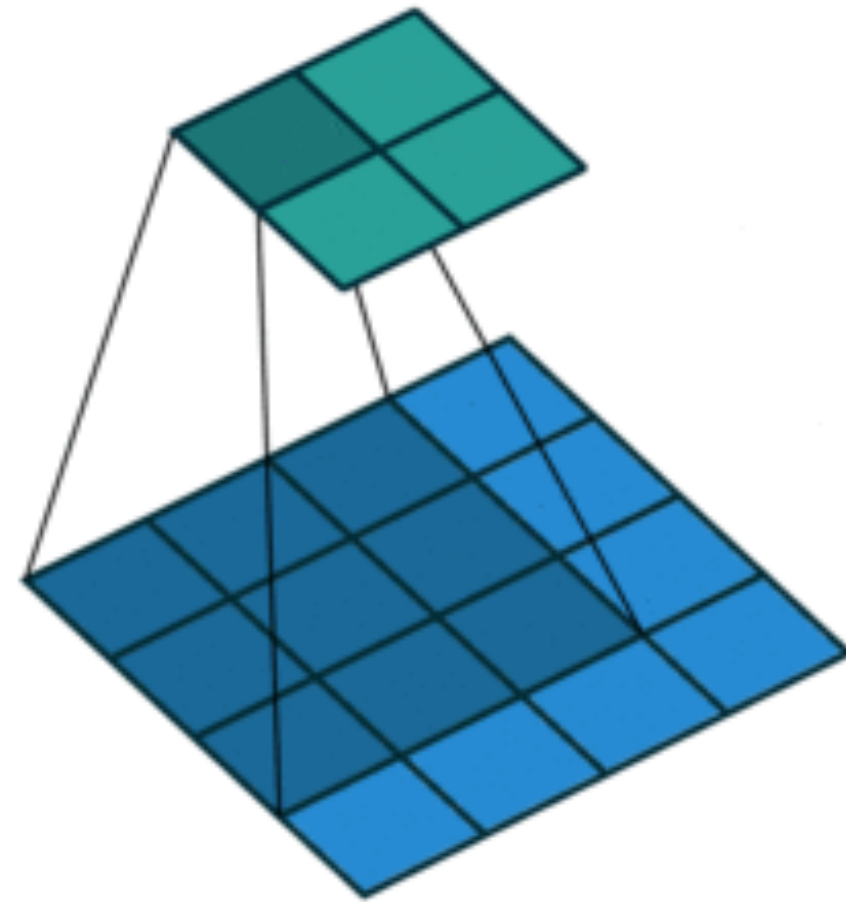


Stride=1, No padding

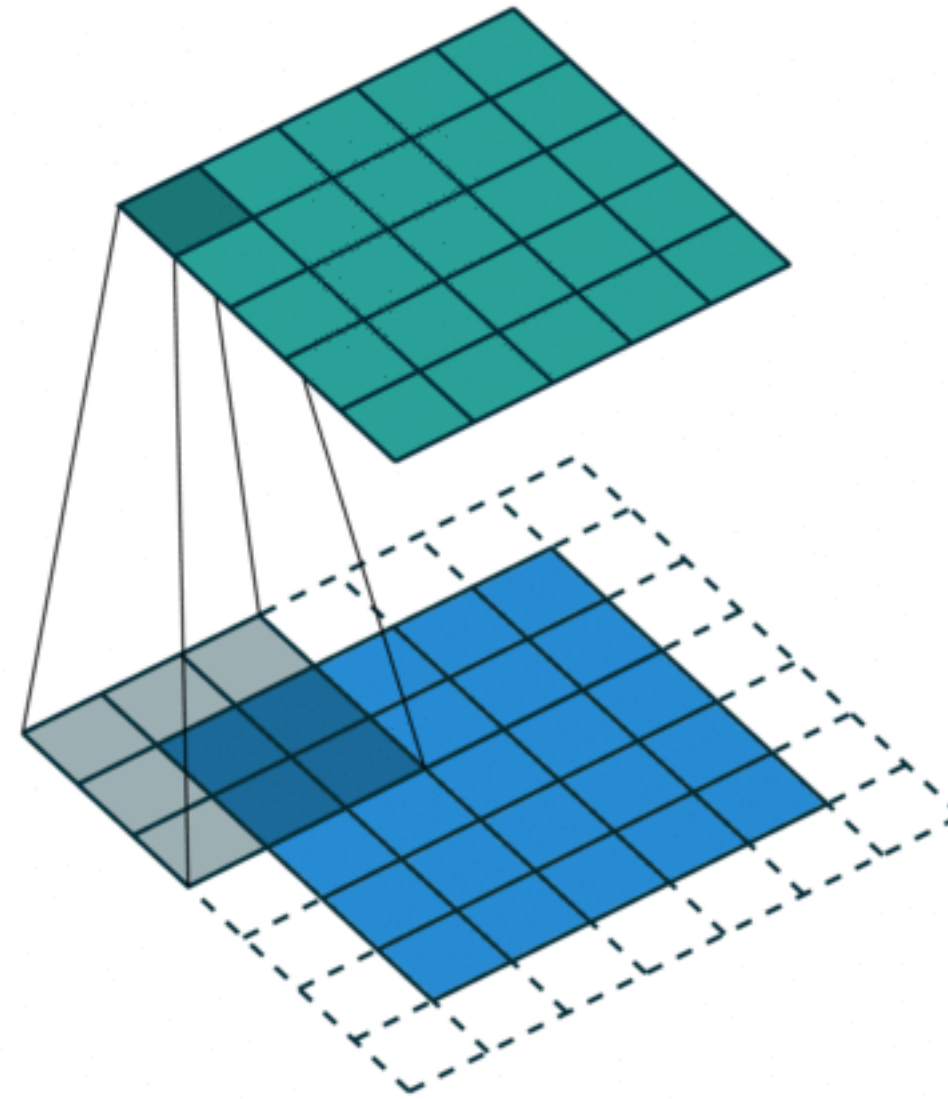


Stride=1, Padding, P=1

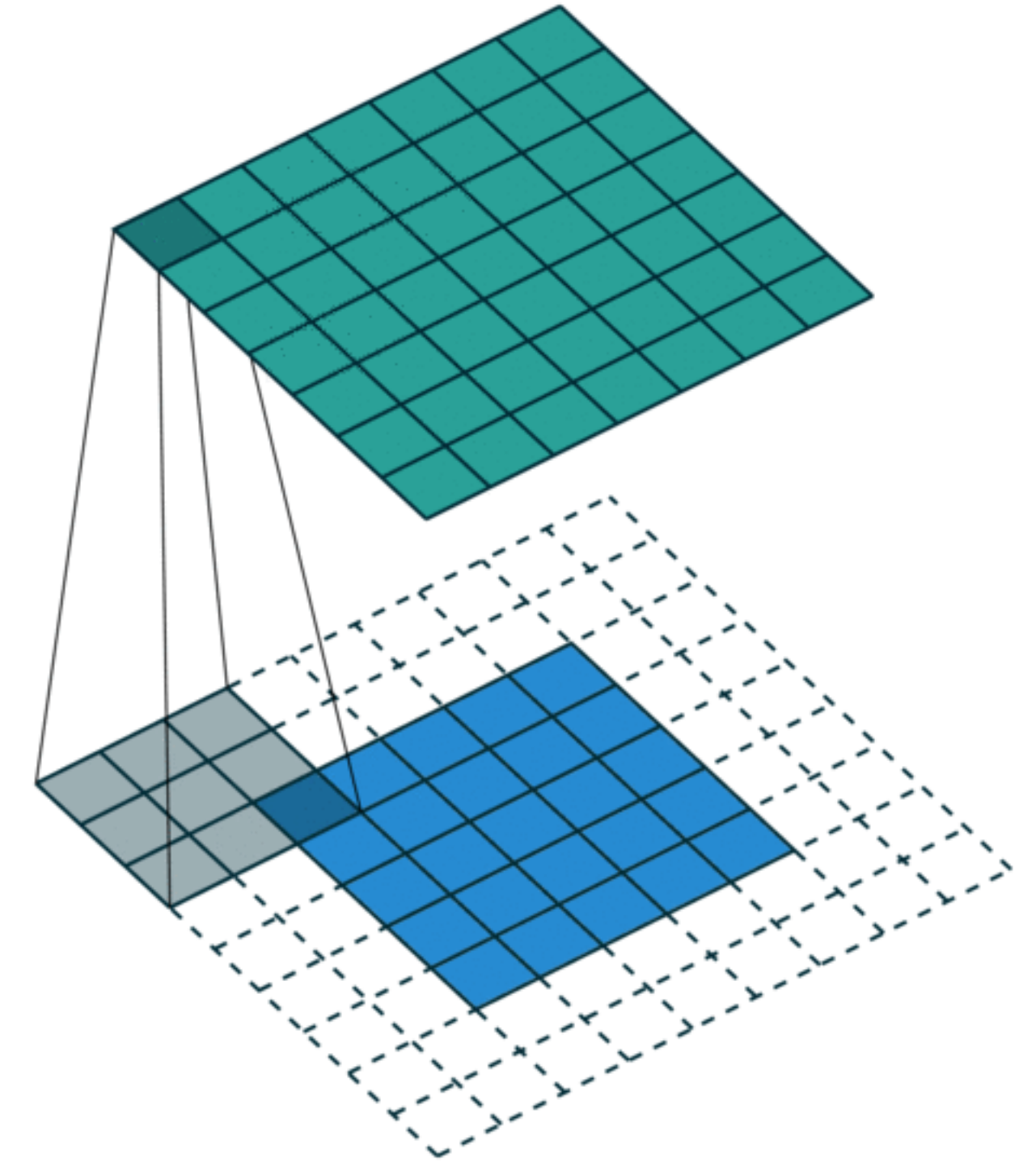
Convolutional Neural Networks (CNNs)



Stride=1, No padding

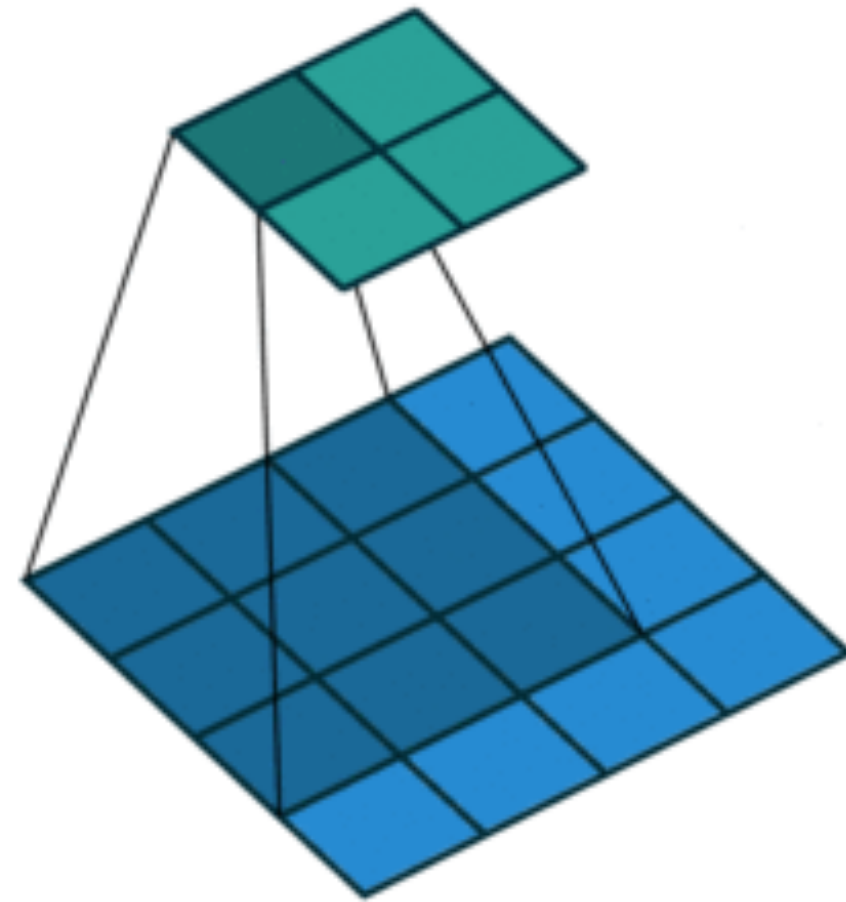


Stride=1, Padding, P=1

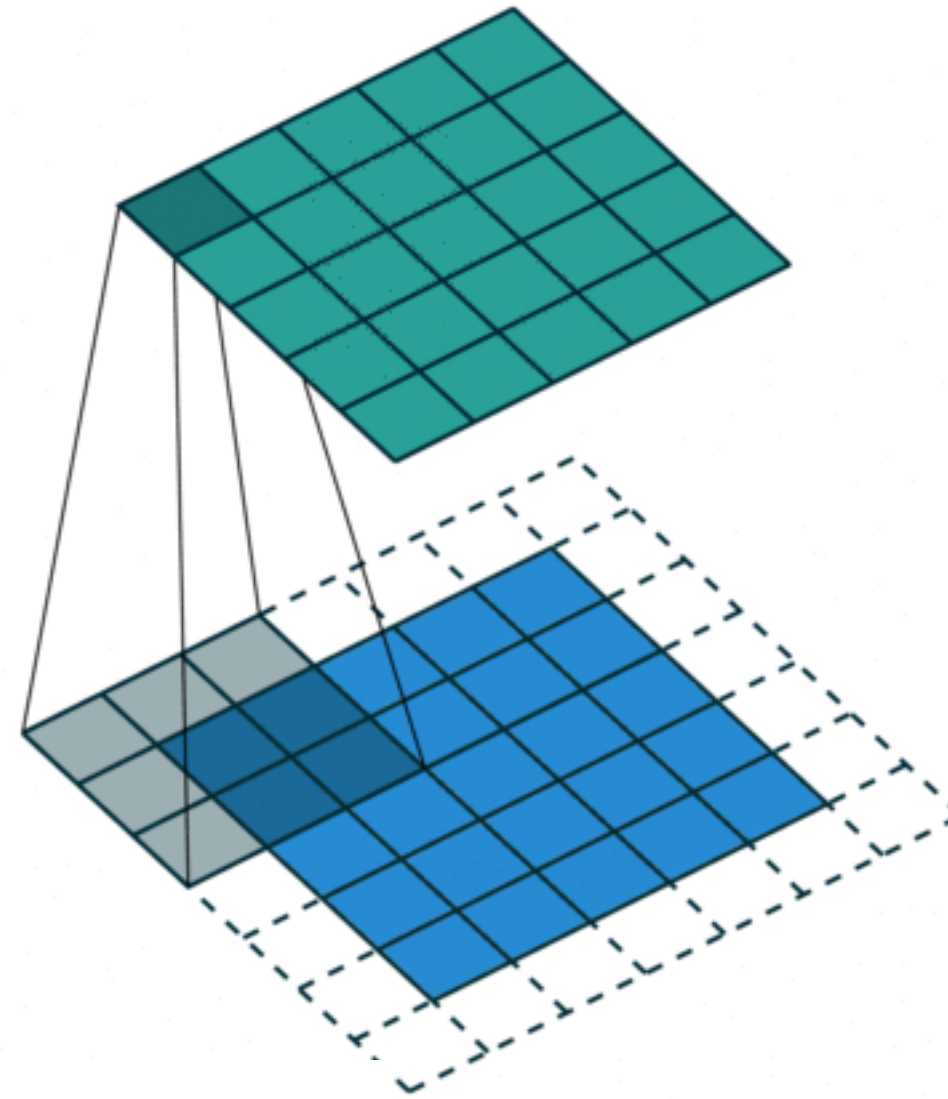


Stride=1, Padding, P=2

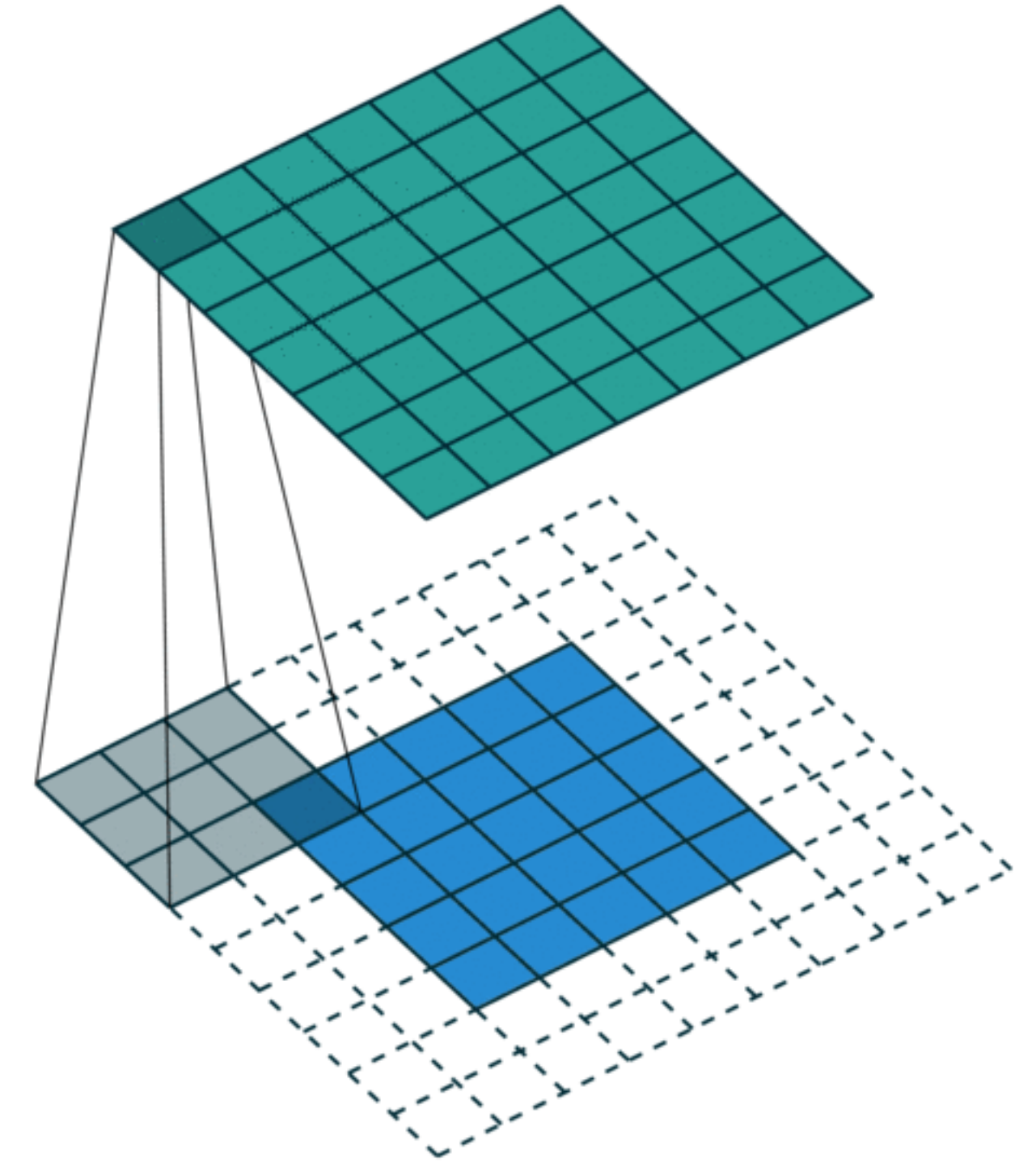
Convolutional Neural Networks (CNNs)



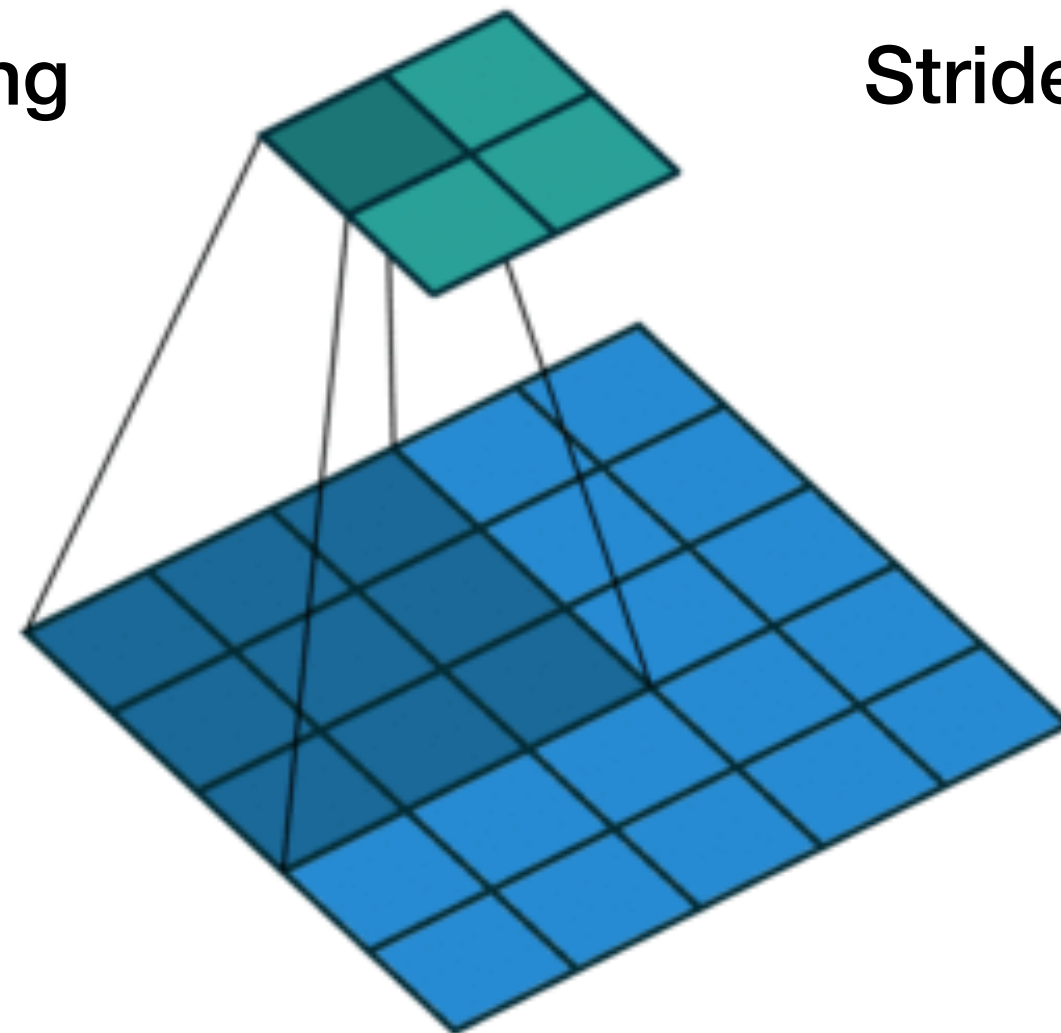
Stride=1, No padding



Stride=1, Padding, P=1

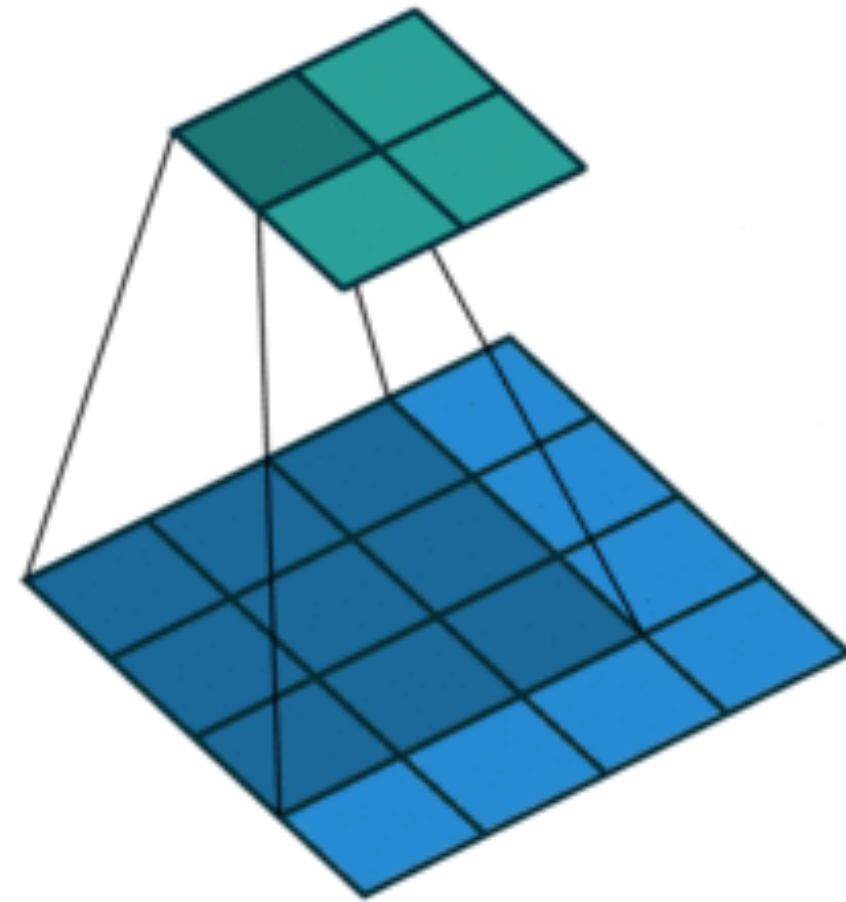


Stride=1, Padding, P=2

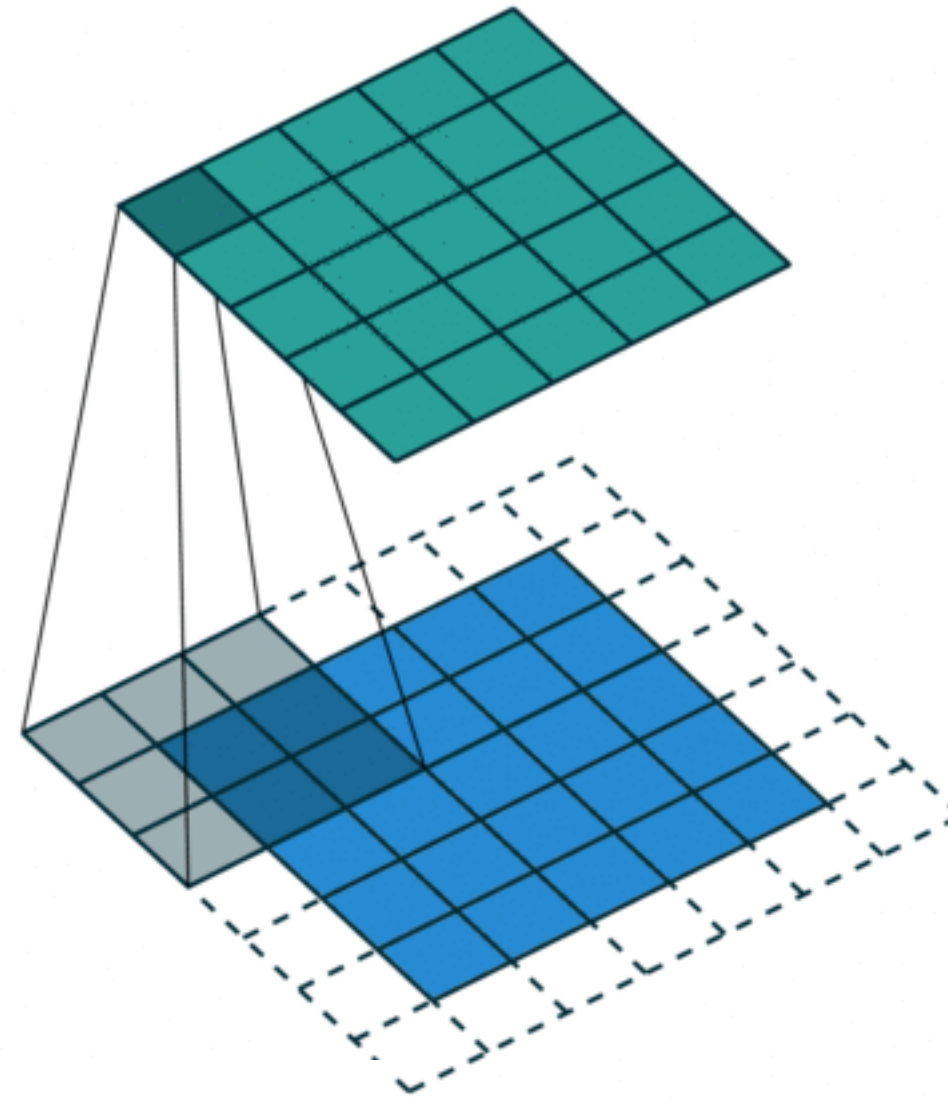


Stride=2, No padding

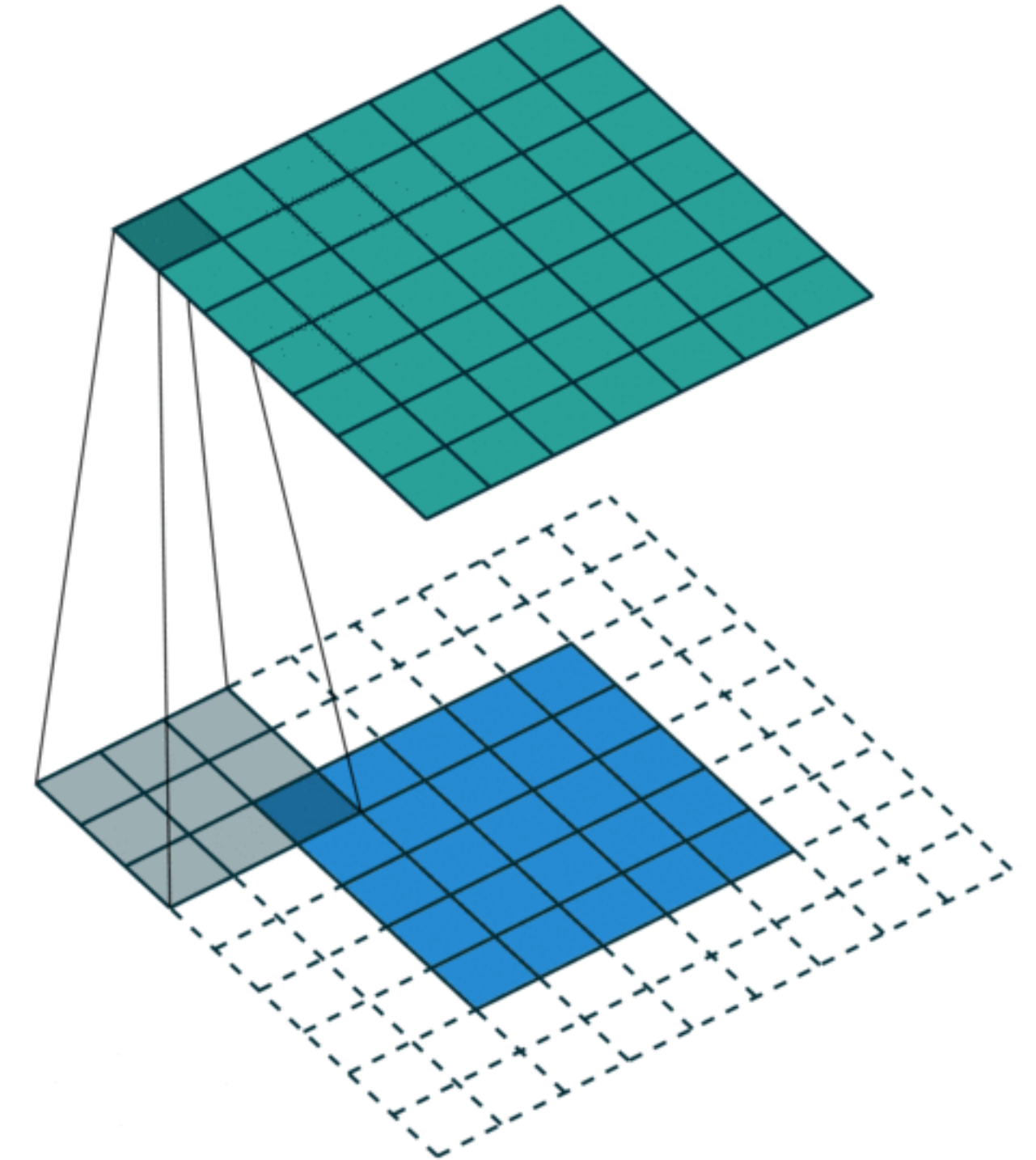
Convolutional Neural Networks (CNNs)



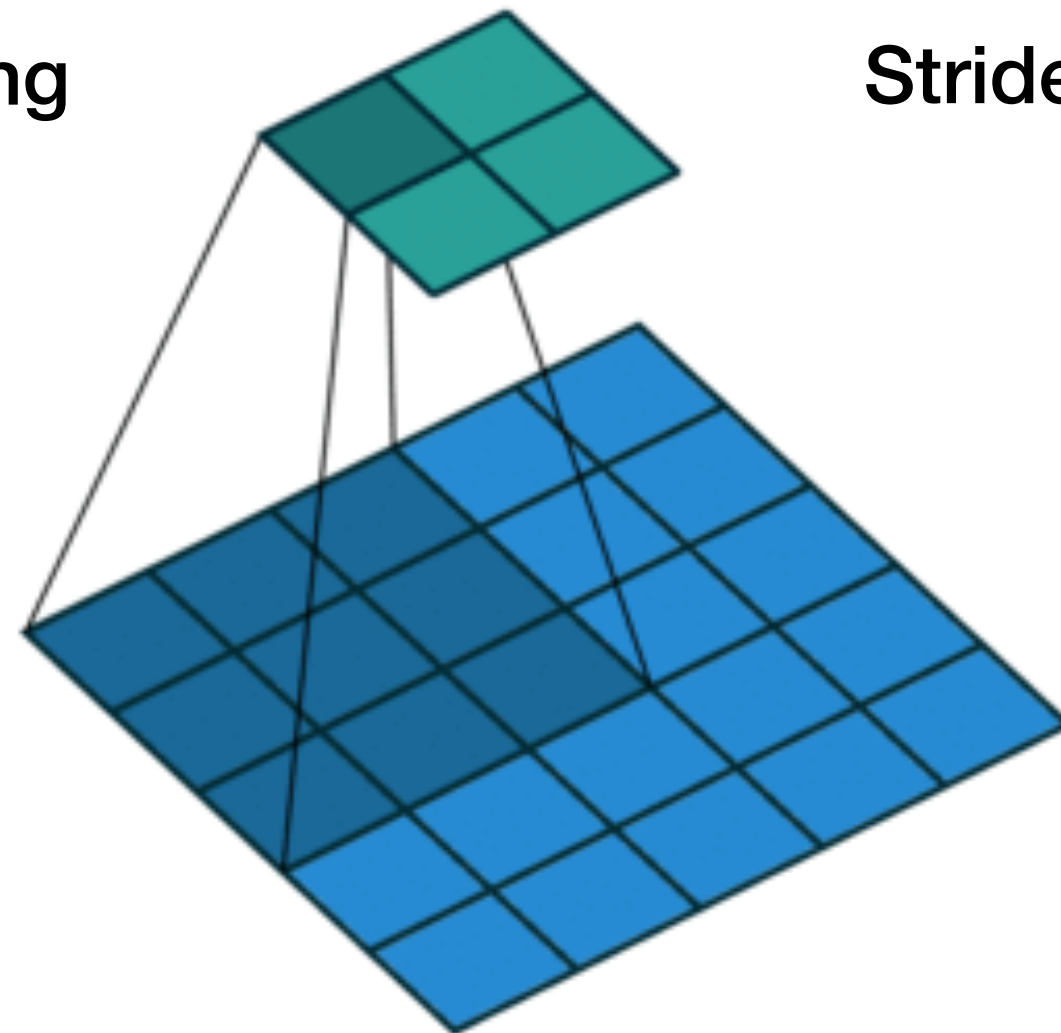
Stride=1, No padding



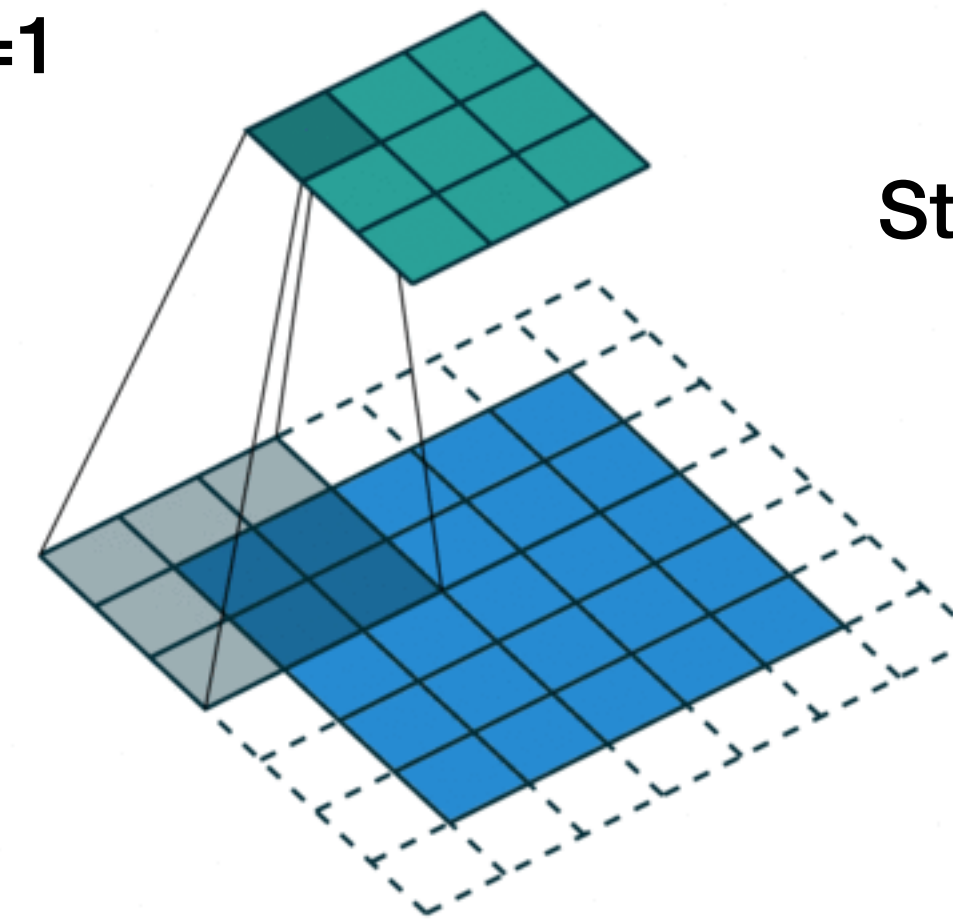
Stride=1, Padding, P=1



Stride=1, Padding, P=2



Stride=2, No padding

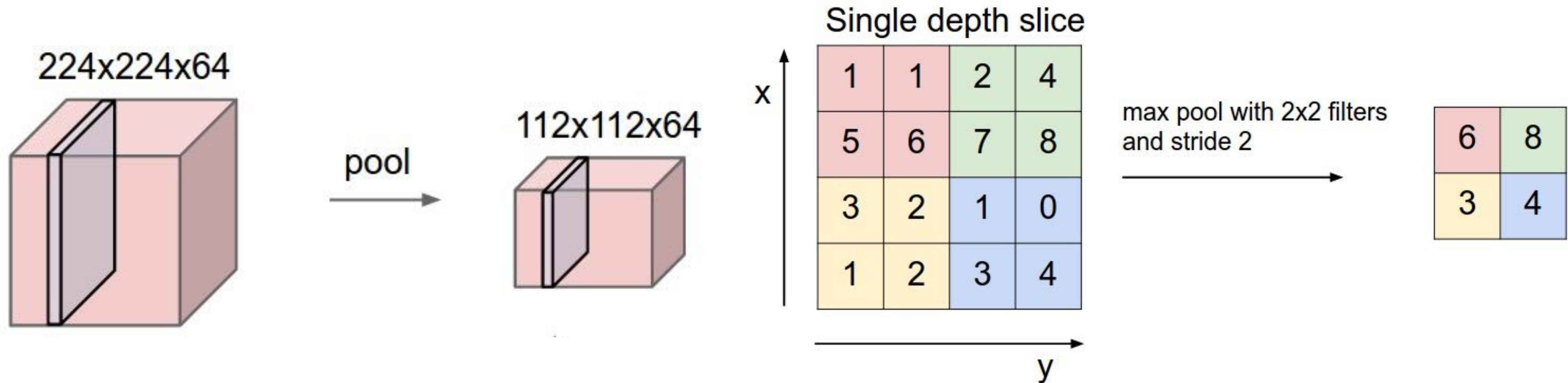


Stride=2, Padding, P=1

Convolution Layers: Summary

- Accepts a volume of size $W_1 \times H_1 \times D_1$
- Requires four hyperparameters:
 - Number of filters K ,
 - their spatial extent F ,
 - the stride S ,
 - the amount of zero padding P .
- Produces a volume of size $W_2 \times H_2 \times D_2$ where:
 - $W_2 = \lfloor (W_1 - F + 2P)/S \rfloor + 1$
 - $H_2 = \lfloor (H_1 - F + 2P)/S \rfloor + 1$ (i.e. width and height are computed equally by symmetry)
 - $D_2 = K$
- With parameter sharing, it introduces $F \cdot F \cdot D_1$ weights per filter, for a total of $(F \cdot F \cdot D_1) \cdot K$ weights and K biases.

Pooling Layer

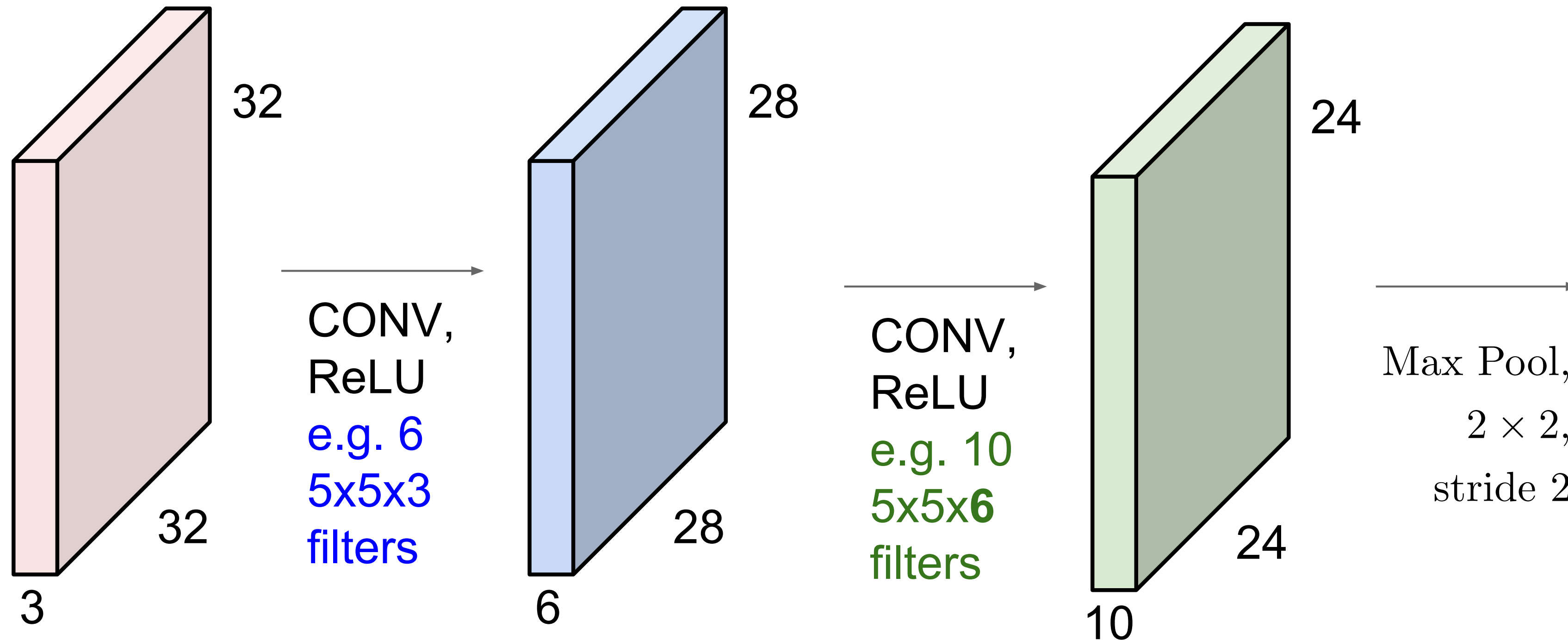


- Why pooling?
Reduce the size of the representation, speed up the computations and make the features a little more robust.
- Max pooling is popularly used in CNNs.

Pooling Layer

- Accepts a volume of size $W_1 \times H_1 \times D_1$
- Requires two hyperparameters:
 - their spatial extent F ,
 - the stride S ,
- Produces a volume of size $W_2 \times H_2 \times D_2$ where:
 - $W_2 = \lfloor (W_1 - F)/S \rfloor + 1$
 - $H_2 = \lfloor (H_1 - F)/S \rfloor + 1$
 - $D_2 = D_1$

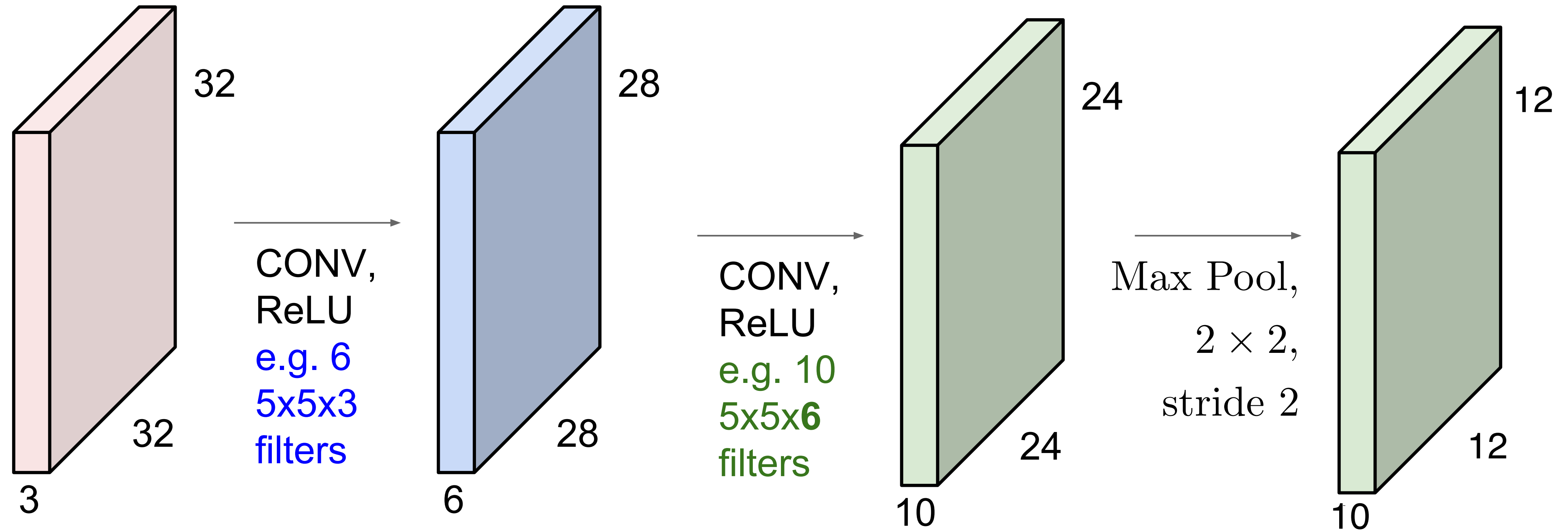
Convolutional NNs



Consider the convolutional neural network shown above with two convolutional layers and a max pooling layer with filters of size 2×2 , stride 2. What are the dimensions of the output feature map?

- ✓ (A) $12 \times 12 \times 10$ (B) $23 \times 23 \times 5$ (C) $11 \times 11 \times 10$ (D) $12 \times 12 \times 5$

Convolutional NNs



Consider the convolutional neural network shown above with two convolutional layers. How many parameters need to be estimated overall for this network? (Ignore biases.)

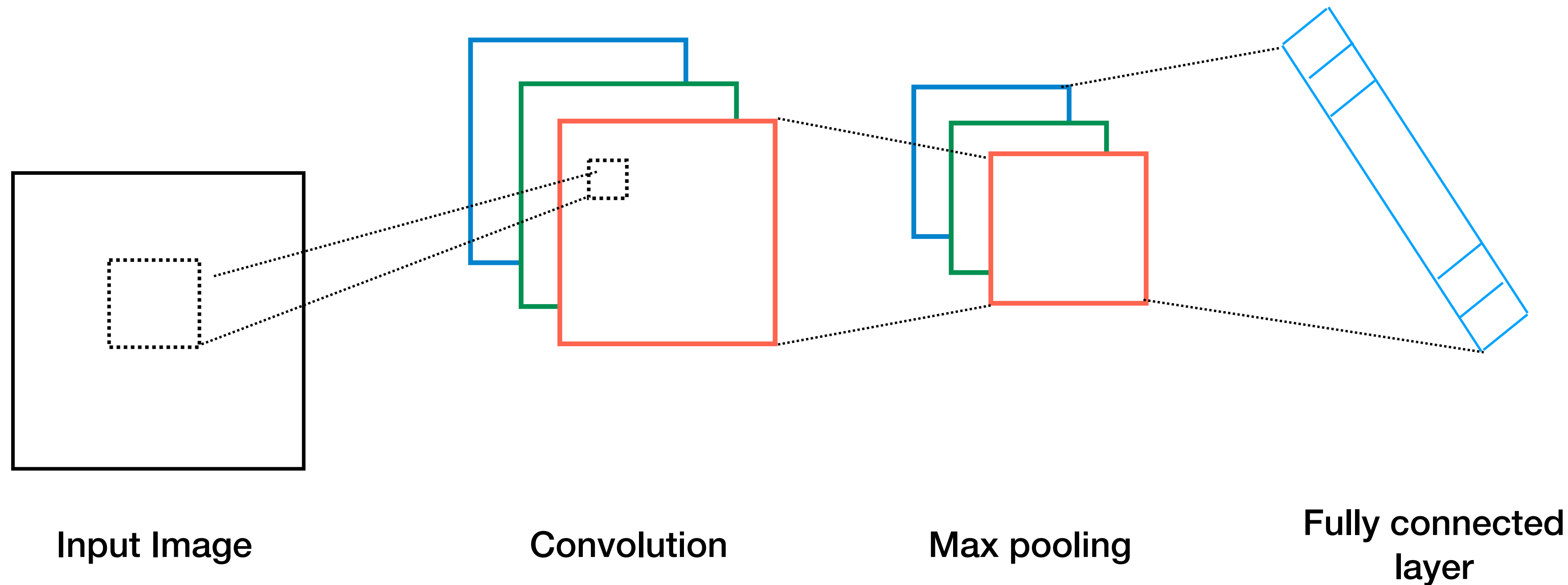
(A) 1500

(B) 1990

✓ (C) 1950

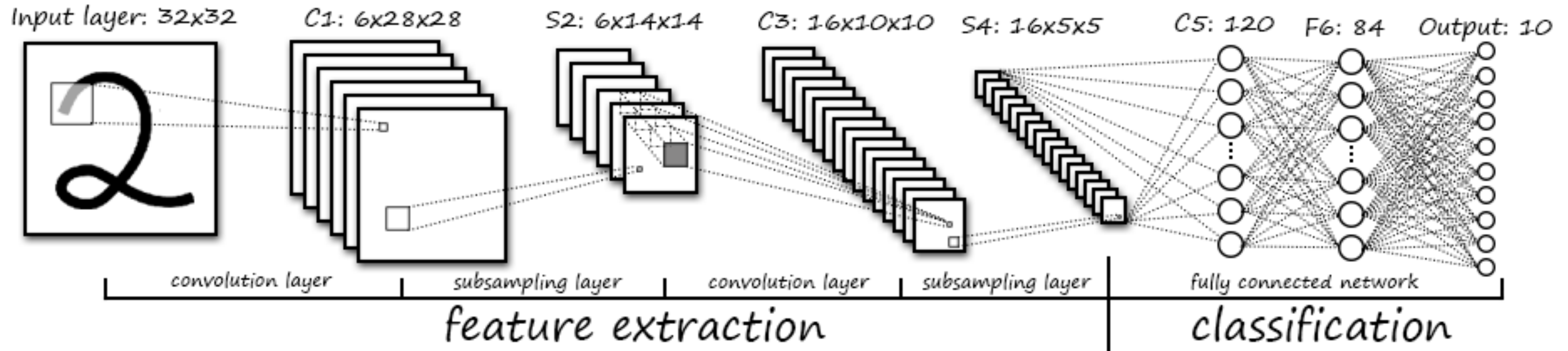
(D) 1954

Convolutional Architectures



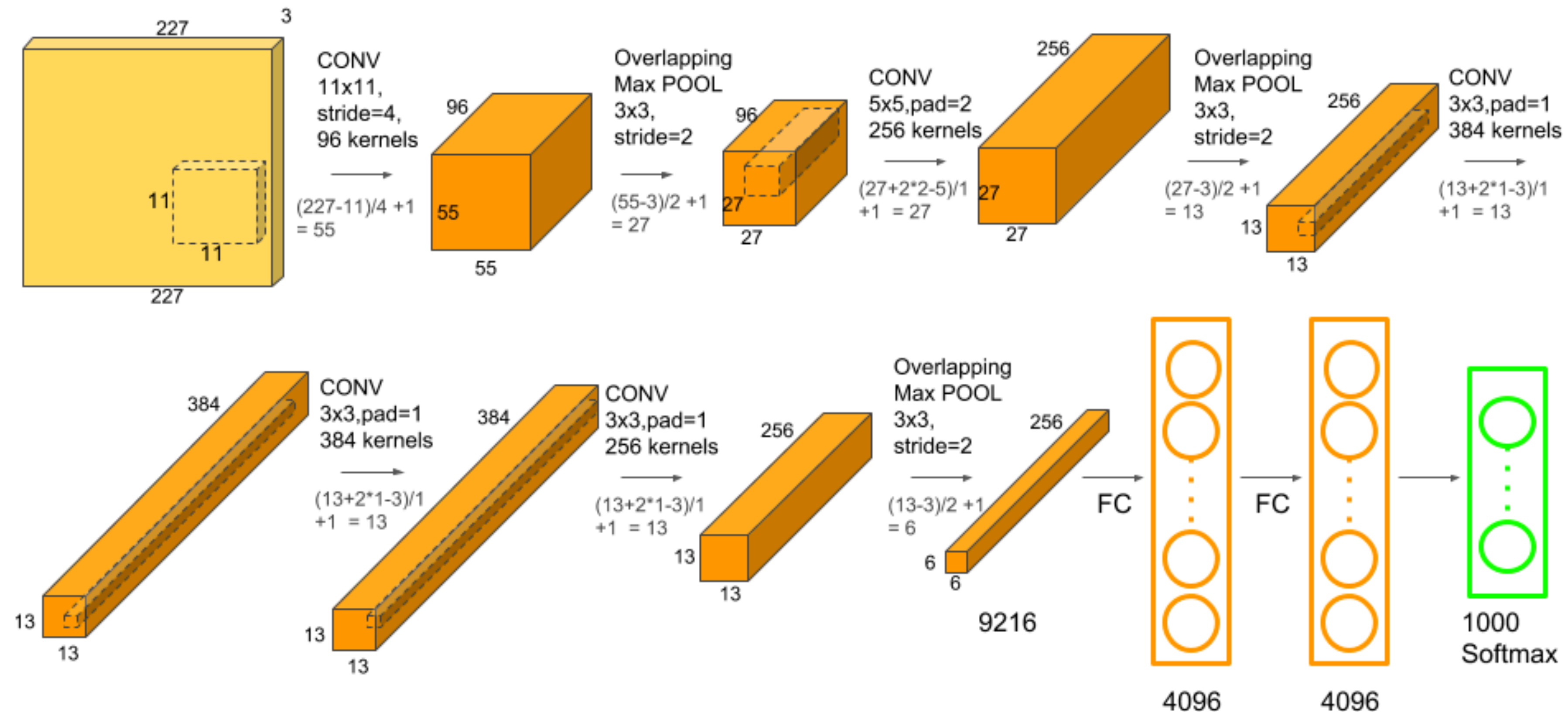
- Block that can be repeated: Convolutional layer, followed by non-linearity (e.g. ReLU) + Max pooling
- Fully connected layers before classification

LeNet-5



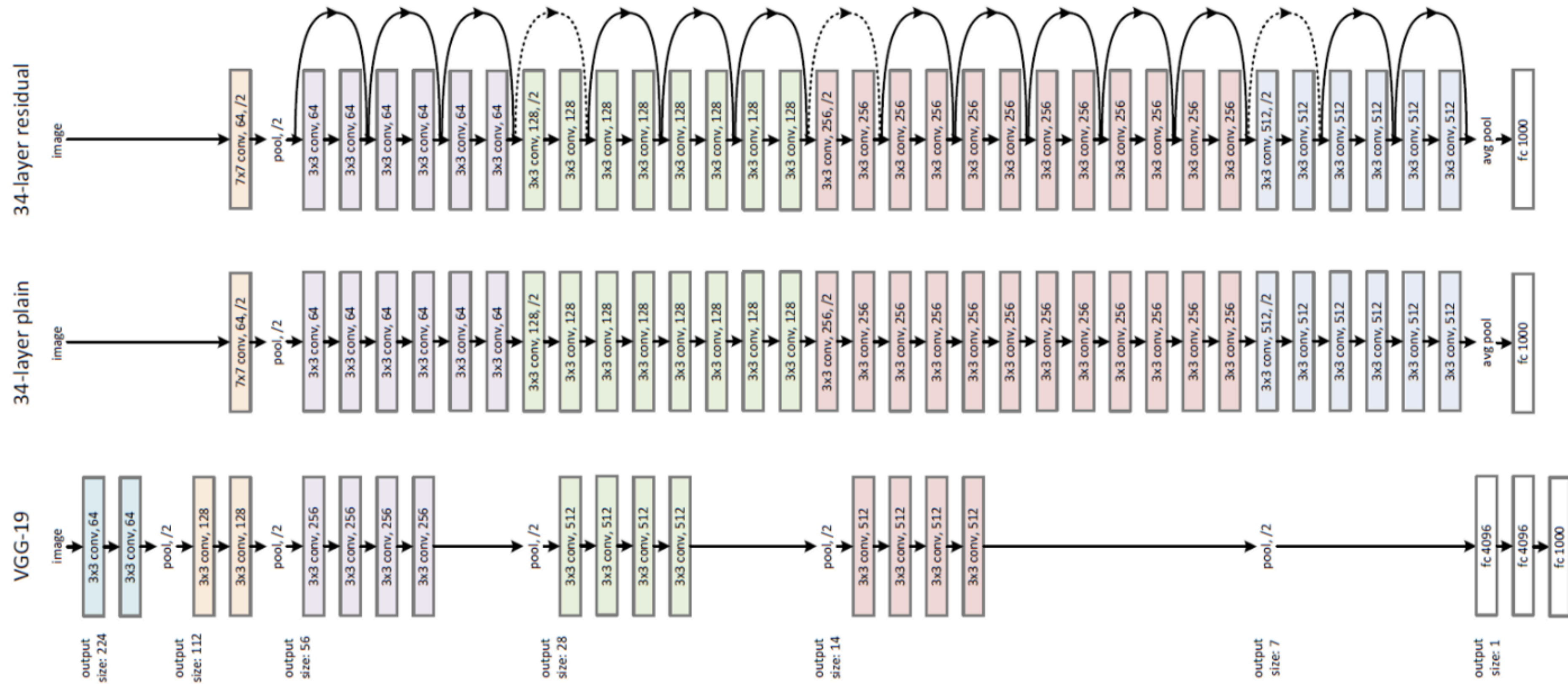
- One of the first successful CNN architectures
- Used to classify images of hand-written digits

AlexNet



- Winner (by a large margin) of the ImageNet challenge in 2012.
- Much larger than previous architectures.

ResNet



- Uses residual blocks