

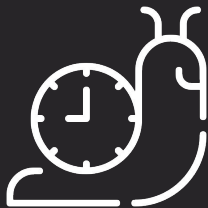


# Module 1: A Beginner's Guide to Computer Vision

## Video 6: Building a CNN - Pooling Layer

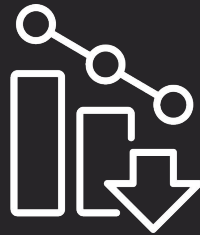
# Problems with Convolution

When working with massive data, convolutions may lead to the following problems:



## Slow Training

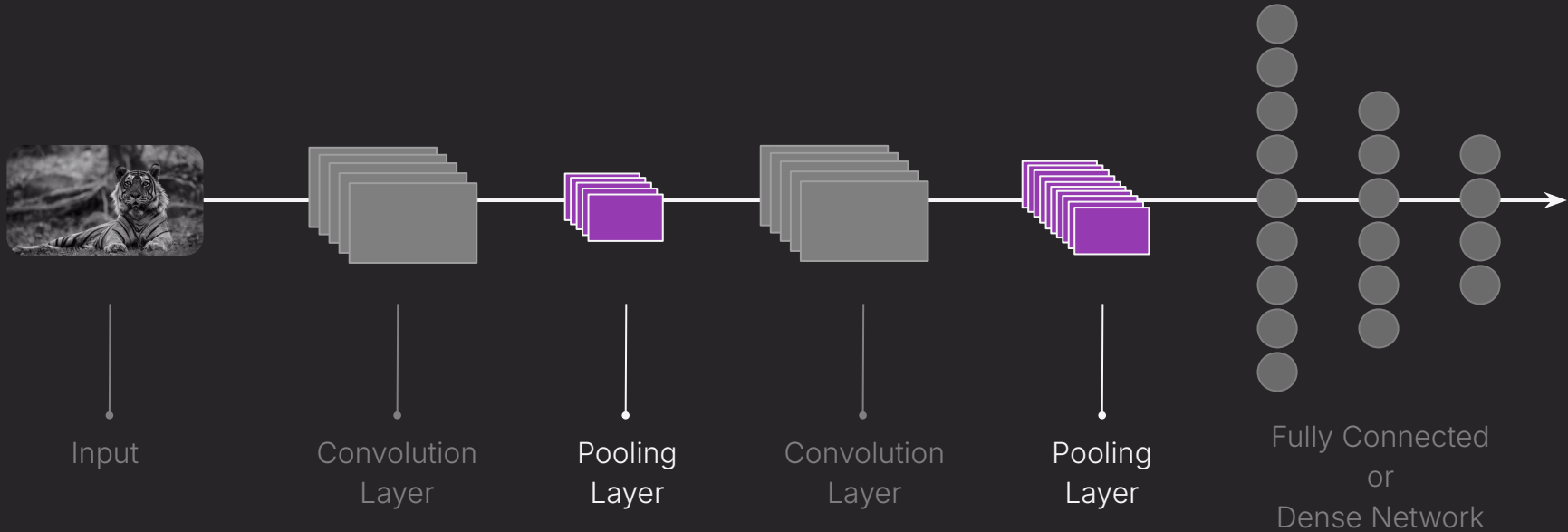
Analyzing every details from a massive database of images will result in long training times.



## Poor Performance

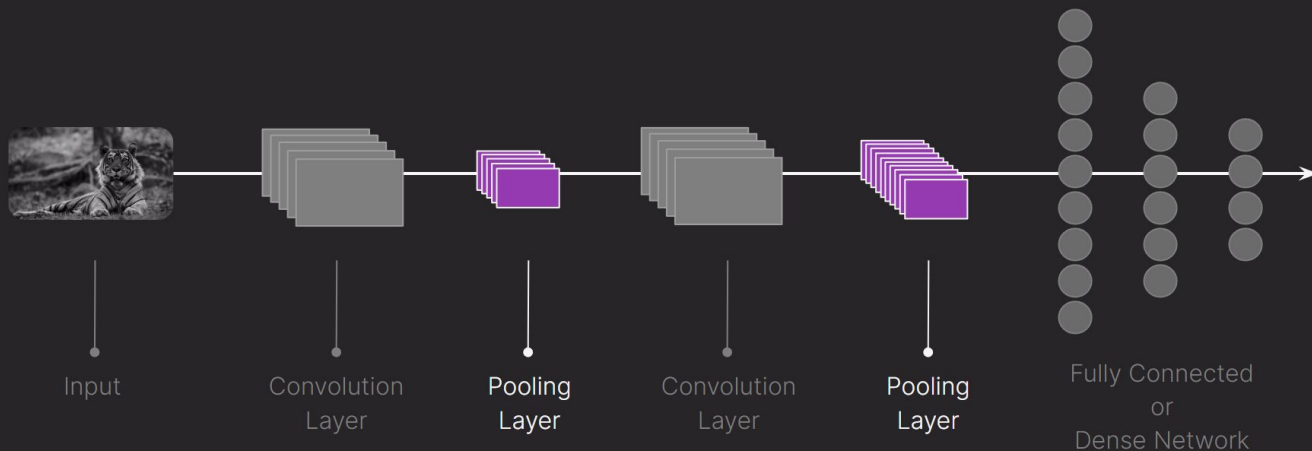
Model might learn irrelevant details during training leading to poor performance on unseen datasets.

# The Pooling Layer



# The Pooling Layer

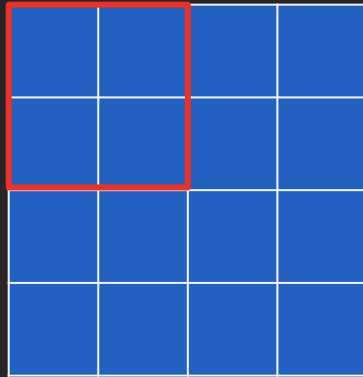
- Reduces complexity of the image data by downsampling the feature maps.
- It focuses on essential details and maintains the overall pattern or structure.



# Types of Pooling

The two main types of pooling are:

- Max Pooling
- Average Pooling



Original 4×4 Feature Map

Applying 2×2 Pooling



Output 2×2 Feature Map

# Types of Pooling: Max Pooling

**Max pooling** takes the maximum value from each path of the feature map.

1	3	2	4
5	6	7	8
9	2	10	1
4	3	2	0

Input 4×4 Feature Map

Applying 2×2 Max Pooling

6	

Output 2×2 Feature Map

# Types of Pooling: Max Pooling

**Max pooling** takes the maximum value from each path of the feature map.

1	3	2	4
5	6	7	8
9	2	10	1
4	3	2	0

Input 4×4 Feature Map

Applying 2×2 Max Pooling

6	8

Output 2×2 Feature Map

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1	3	2	4
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Input 4×4 Feature Map

Applying 2×2 Max Pooling

6	8
9	

Output 2×2 Feature Map



# Types of Pooling: Max Pooling

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

Input 4×4 Feature Map

Applying 2×2 Max Pooling

6	8
9	10

Output 2×2 Feature Map

# Types of Pooling: Max Pooling



**Input Image**



**Max Pooled Image**

# Types of Pooling: Average Pooling

**Average pooling** takes the average value from each path of the feature map.

1	3	2	4
5	6	7	8
9	2	10	1
4	3	2	0

Input 4×4 Feature Map

$$= (1+3+5+6)/4 = 3.75$$

Rounded down to 3

3	

Output 2×2 Feature Map

# Types of Pooling: Average Pooling

**Average pooling** takes the average value from each path of the feature map.

1	3	2	4
5	6	7	8
9	2	10	1
4	3	2	0

Input 4×4 Feature Map

$$= (2+4+7+8)/4=5.25$$

Rounded down to 5

3	5

Output 2×2 Feature Map

# Types of Pooling: Average Pooling

**Average pooling** takes the average value from each path of the feature map.

1	3	2	4
5	6	7	8
9	2	10	1
4	3	2	0

Input 4×4 Feature Map

$$= (10+1+2+0)/4=3.25$$

Rounded down to 3

3	5
4	<b>3</b>

Output 2×2 Feature Map

# Types of Pooling: Average Pooling



**Input Image**



**Average Pooled Image**

# Pooling Formula

1	3	2	4
5	6	7	8
9	2	10	1
4	3	2	0

Input 4×4 Feature Map

3	5
4	<b>3</b>

Output 2×2 Feature Map

$$\text{Output size} = \frac{W - P_s}{S} + 1$$

- W = Width of the input
- $P_s$  = Pooling Size
- S = Stride used

# Pooling Formula

1	3	2	4
5	6	7	8
9	2	10	1
4	3	2	0

Input 4×4 Feature Map

3	5
4	<b>3</b>

Output 2×2 Feature Map

$$\text{Output size} = \frac{W - P_s}{S} + 1$$

$$\begin{aligned}\text{Output size} &= \frac{4 - 2}{2} + 1 \\ &= 2\end{aligned}$$



# The Pooling Layer



Recognize objects regardless of their position or orientation.



Contributes to the model's generalization ability.

**Max Pooled Image**



**Up Next : Module 2: Building Blocks for Image Recognition**