



MODERN COMPUTER VISION

BY RAJEEV RATAN

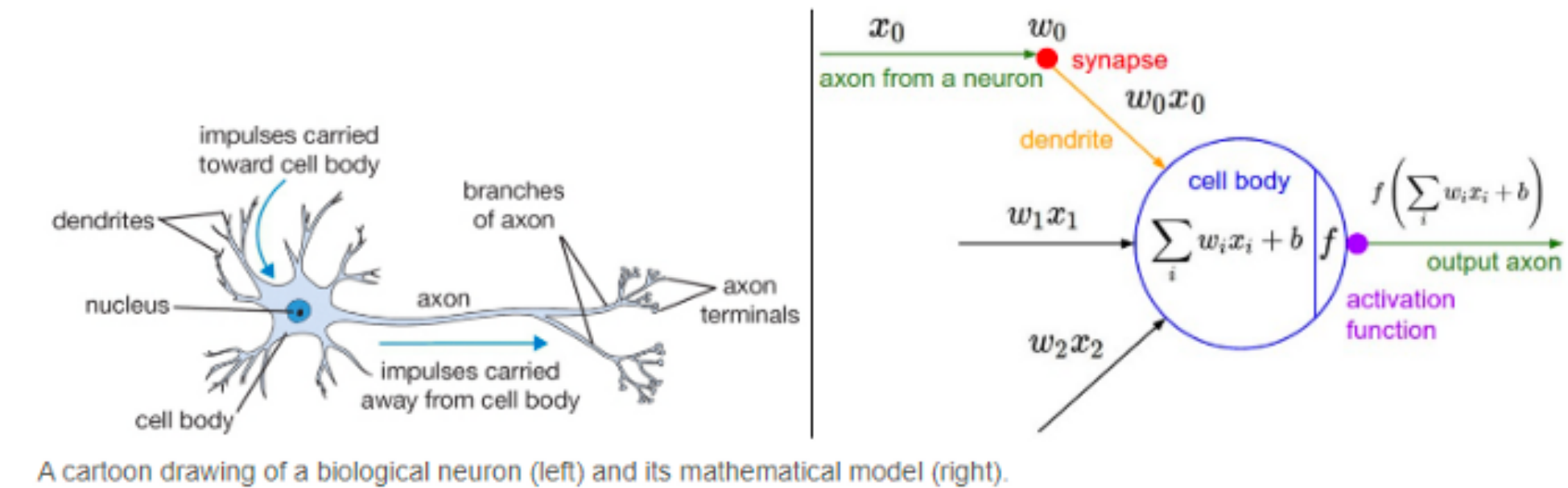
Activation Layer ReLU

What are Activation Functions and their importance

Purpose of Activation Functions

To enable the learning of **complex patterns** in our data

- Biological neurons fire (activate) on certain inputs, these are then fed into other neurons
- Introduces **non-linearity** to our network
- This allows a non-linear decision boundary via non-linear combinations of the weight and inputs



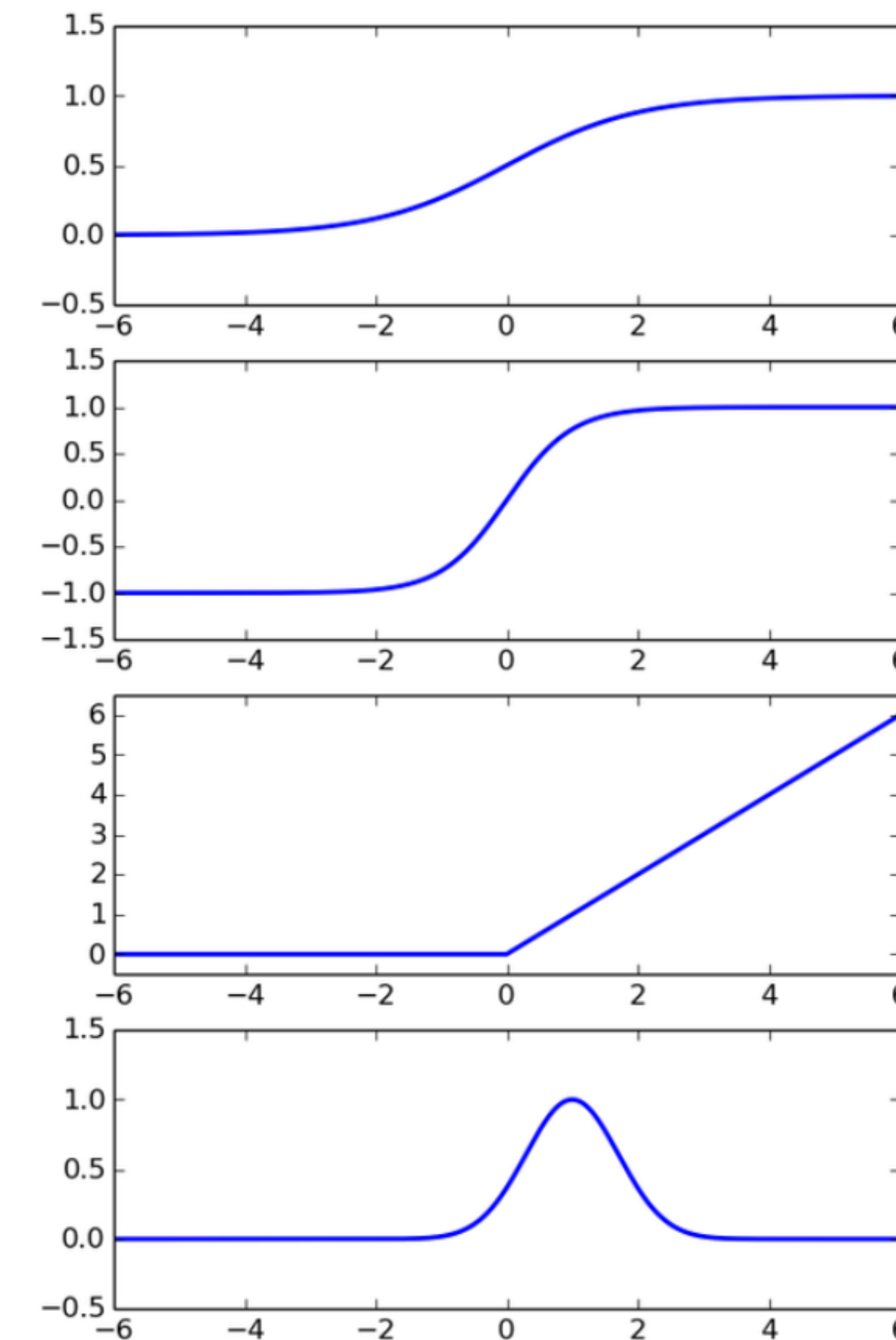
Stanford's CS231 Course

Types of Activation Functions

There are several activation functions we can use in our CNN. However, Rectified Linear Units (**ReLU**) have become the activation function of choice for CNNs.

ReLU is advantageous in CNN Training:

- Simple Computation (fast to train)
- Does not saturate



Sigmoid

$$\phi(z) = \frac{1}{1 + e^{-z}}$$

Hyperbolic Tangent

$$\phi(z) = \frac{e^z - e^{-z}}{e^z + e^{-z}}$$

Rectified Linear

$$\phi(z) = \begin{cases} 0 & \text{if } z < 0 \\ z & \text{if } z \geq 0 \end{cases}$$

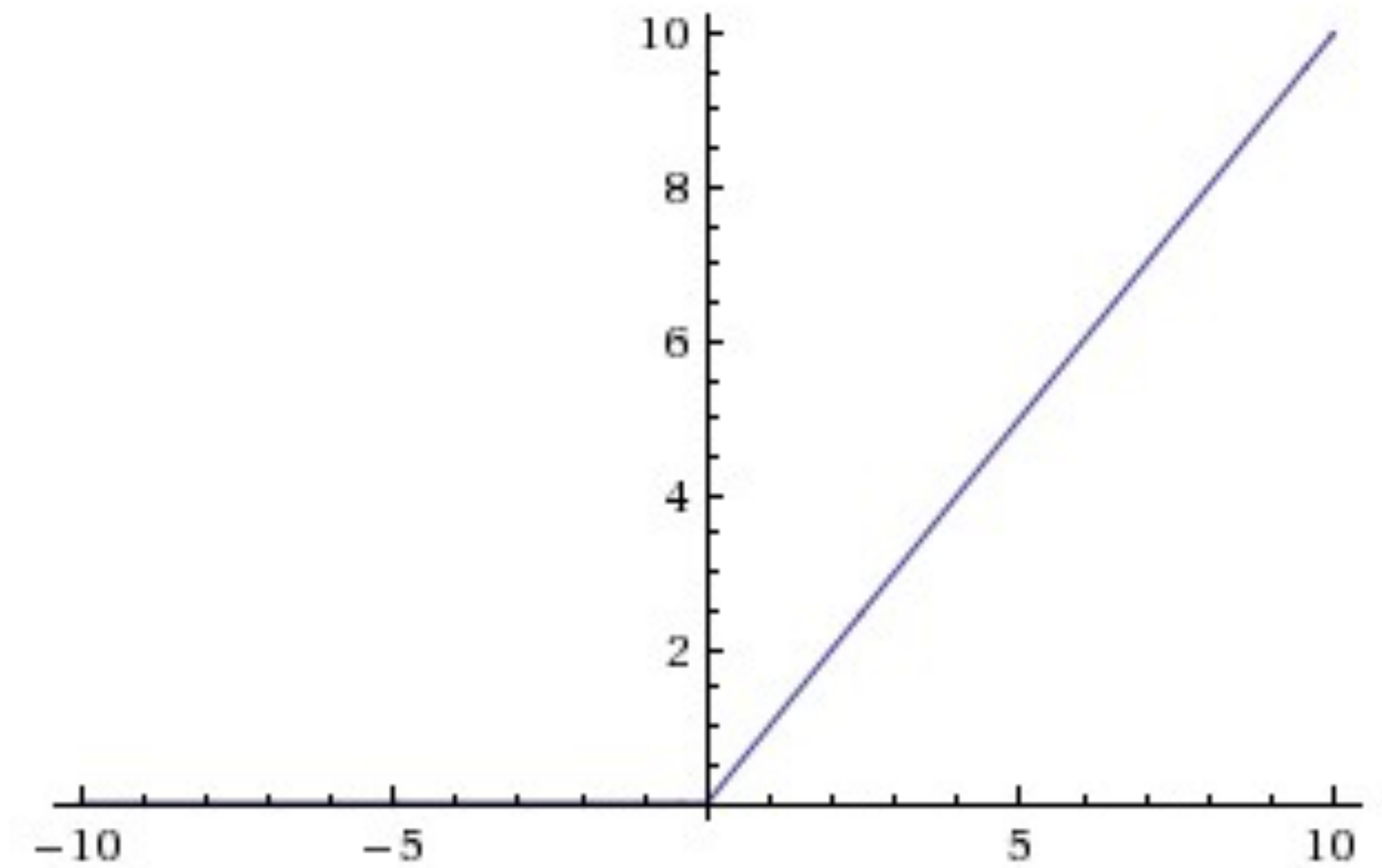
Radial Basis Function

$$\phi(z, c) = e^{-(\epsilon \|z - c\|)^2}$$

The ReLU Operation

- Change all negative values to 0
- Leave all positive Values alone

$$f(x) = \max(0, x)$$



Applying the ReLU Activation

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

Input Image

*

0	1	0
1	0	-1
0	1	0

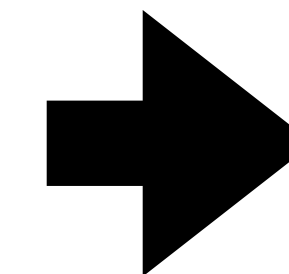
Filter or Kernel

=

2	1	-1
-1	1	3
2	1	-5

Output or Feature Map

ReLU



2	1	0
0	1	3
2	1	0

Applying the ReLU Activation

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

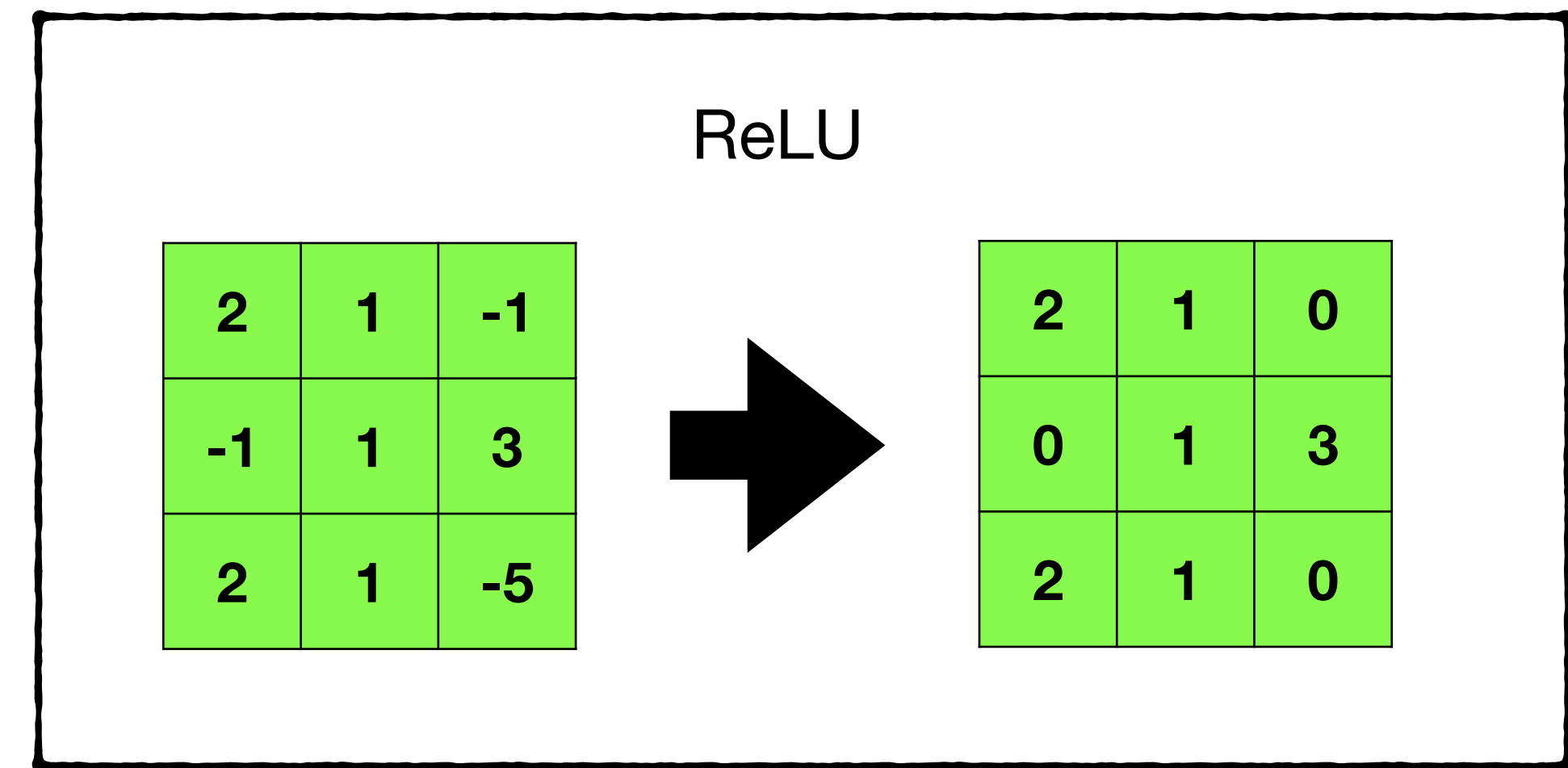
Input Image

*

0	1	0
1	0	-1
0	1	0

Filter or Kernel

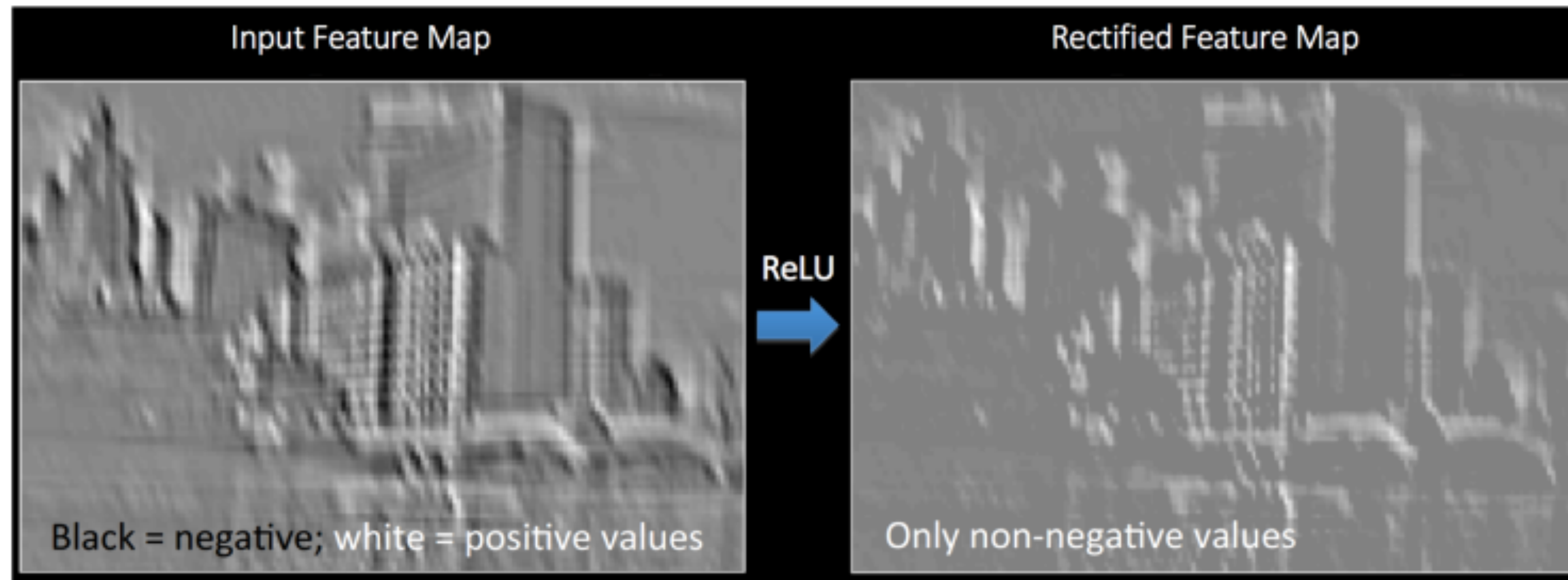
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Output or Feature Map

Rectified Feature Map

Example of a Rectified Linear Map



Source - http://mlss.tuebingen.mpg.de/2015/slides/fergus/Fergus_1.pdf



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Next...

Pooling