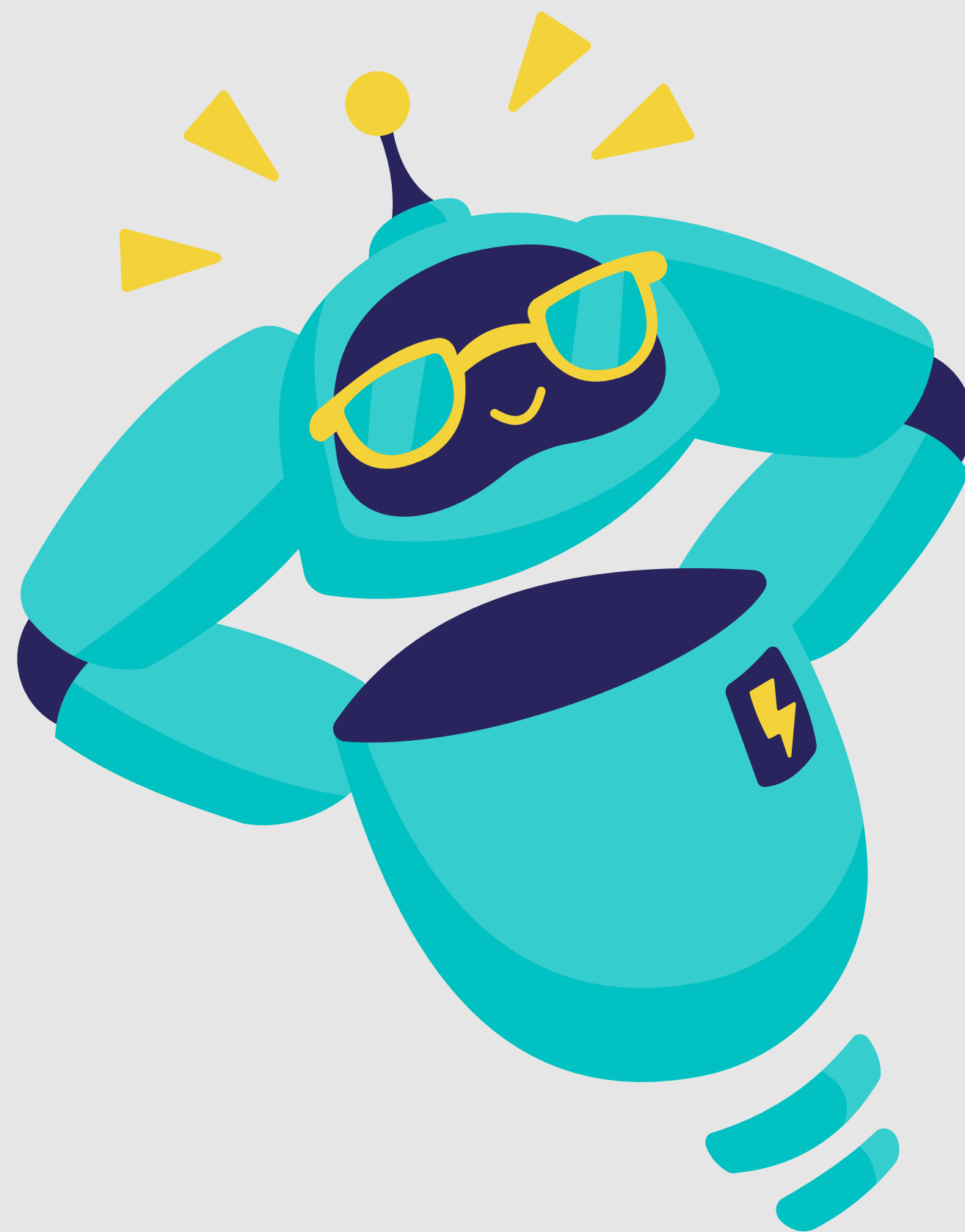




DEEPFLOW

CNN WORKSHOP

HELD BY YUCEF OUHAB





DEEPFLOW

CONVOLUTIONAL NEURAL NETWORKS (CNNs):

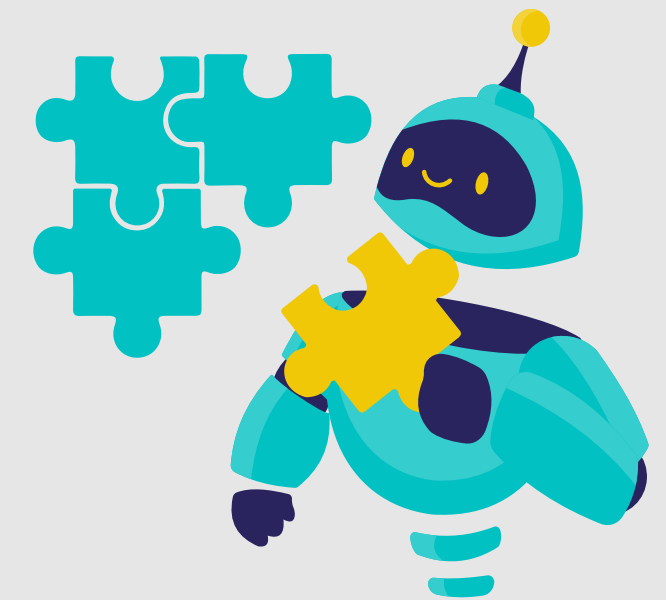
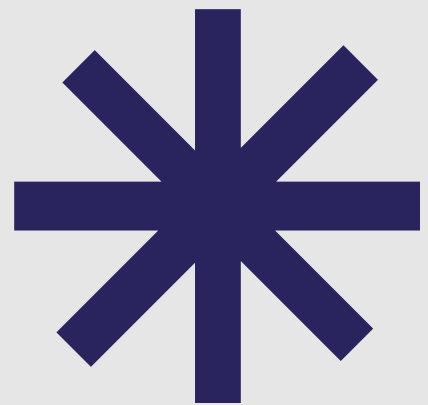
A DEEP DIVE

DEFINITION OF CNN



DEEPFLOW

A CNN IS A TYPE OF DEEP LEARNING MODEL DESIGNED FOR PROCESSING STRUCTURED GRID DATA, SUCH AS IMAGES. IT AUTOMATICALLY AND ADAPTIVELY LEARNS SPATIAL HIERARCHIES OF FEATURES FROM INPUT IMAGES.

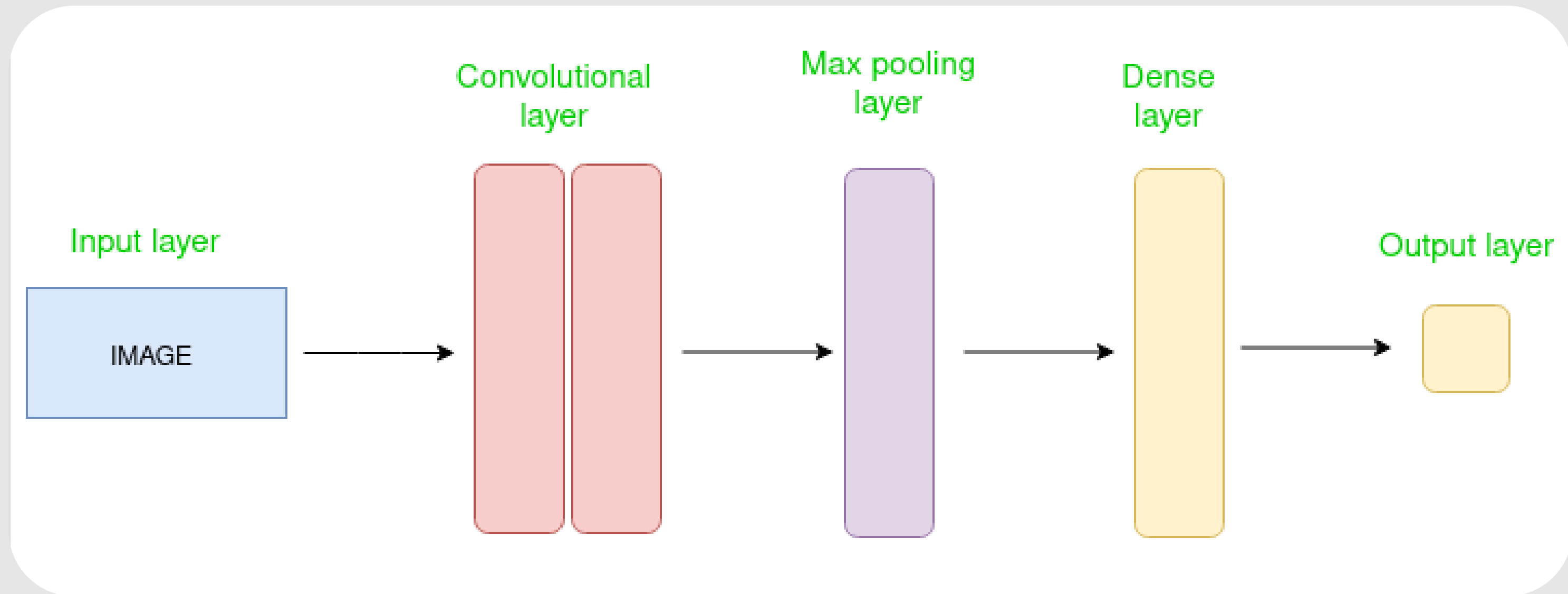


source:<https://www.geeksforgeeks.org/introduction-convolution-neural-network/>

DEFINITION OF CNN



DEEPFLOW



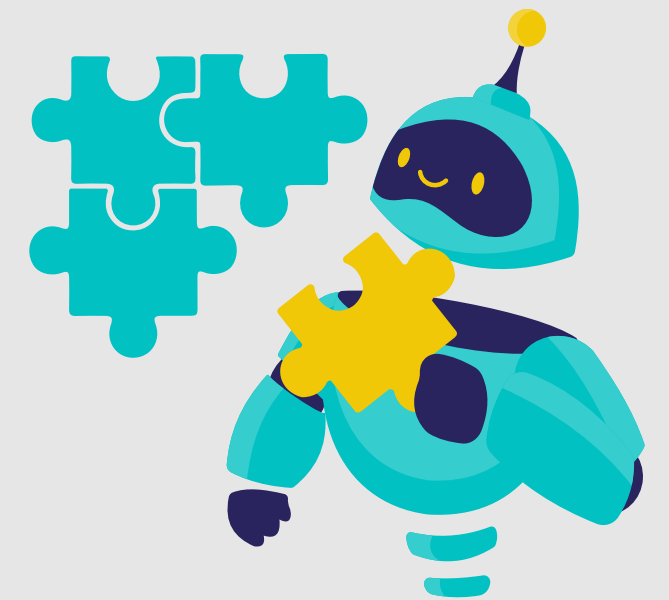
source:<https://www.geeksforgeeks.org/introduction-convolution-neural-network/>

DEFINITION OF CNN



DEEPFLOW

CNNS EXCEL IN TASKS INVOLVING IMAGE RECOGNITION AND PROCESSING
DUE TO THEIR ABILITY TO CAPTURE SPATIAL FEATURES.

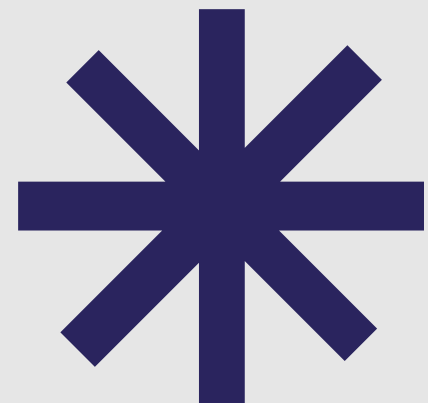


DEFINITION OF CNN



DEEPFLOW

THE ARCHITECTURE
TYPICALLY INCLUDES:



CONVOLUTIONAL LAYERS
FOR FEATURE EXTRACTION

POOLING LAYERS FOR
DIMENSIONALITY
REDUCTION

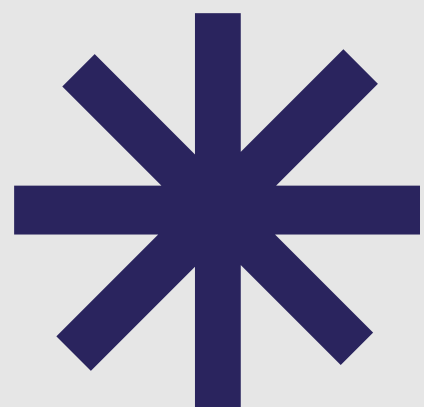
FULLY CONNECTED (FC)
LAYERS FOR
CLASSIFICATION.

WHY CNNs?



DEEPFLOW

CNNs CAPTURE LOCAL PATTERNS IN INITIAL LAYERS AND MORE
GLOBAL PATTERNS IN DEEPER LAYERS

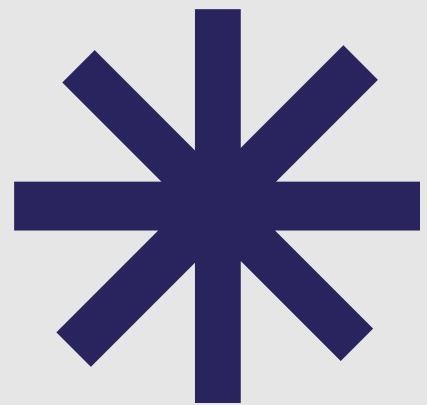


WHY CNNs?



DEEPFLOW

THE DESIGN OF CNNs IS INFLUENCED BY THE HUMAN VISUAL SYSTEM,
PARTICULARLY THE WAY NEURONS RESPOND TO OVERLAPPING
REGIONS IN THE VISUAL FIELD.



WHY CNNs?



DEEPFLOW

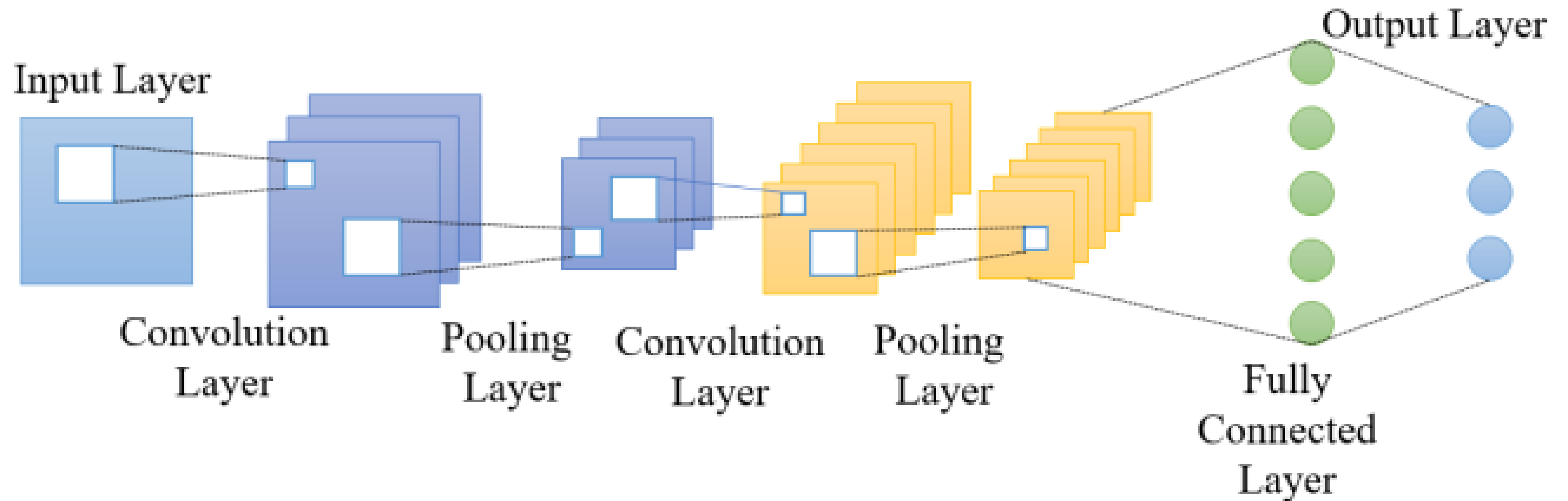
CNNs ARE WIDELY USED IN VARIOUS APPLICATIONS, INCLUDING
MEDICAL IMAGE ANALYSIS, AUTONOMOUS VEHICLES, AND
SURVEILLANCE SYSTEMS.



CNN ARCHITECTURE OVERVIEW



DEEPFLOW



CNN ARCHITECTURE OVERVIEW



DEEPFLOW

TYPICAL SEQUENCE IN A CNN WHERE THE INPUT IMAGE IS PROCESSED THROUGH CONVOLUTIONAL LAYERS (CONV), ACTIVATION FUNCTIONS LIKE RELU, POOLING LAYERS (POOL), FULLY CONNECTED LAYERS (FC), AND FINALLY PRODUCES AN OUTPUT.



INPUT LAYER



DEEPFLOW

INPUT FORMAT (RGB, GRAYSCALE):

IMAGES CAN BE IN COLOR (RGB) WITH THREE CHANNELS OR GRAYSCALE WITH A SINGLE CHANNEL.

SHAPE OF IMAGE DATA:

DEFINED BY DIMENSIONS (HEIGHT, WIDTH) AND THE NUMBER OF CHANNELS (DEPTH).

NORMALIZATION AND PREPROCESSING:

TECHNIQUES LIKE SCALING PIXEL VALUES TO A $[0,1]$ RANGE TO IMPROVE TRAINING EFFICIENCY AND PERFORMANCE.

CONVOLUTIONAL LAYER



DEEPFLOW

WHAT IS A CONVOLUTION OPERATION?

- IN CONVOLUTIONAL NEURAL NETWORKS (CNNs), A CONVOLUTION OPERATION INVOLVES A SMALL MATRIX CALLED A FILTER OR KERNEL SLIDING OVER THE INPUT IMAGE TO PRODUCE A FEATURE MAP. THIS PROCESS HELPS IN DETECTING SPECIFIC FEATURES LIKE EDGES, TEXTURES, OR PATTERNS WITHIN THE IMAGE.

CONVOLUTIONAL LAYER



DEEPFLOW

FILTERS/KERNELS, STRIDE, PADDING:

- FILTERS ARE SMALL MATRICES THAT DETECT SPECIFIC FEATURES
- STRIDE DETERMINES THE MOVEMENT OF THE FILTER
- PADDING INVOLVES ADDING EXTRA PIXELS AROUND THE INPUT TO CONTROL OUTPUT DIMENSIONS.

CONVOLUTIONAL LAYER



DEEPFLOW

PADDING:

PADDING INVOLVES ADDING EXTRA PIXELS (USUALLY ZEROS) AROUND THE BORDER OF THE INPUT IMAGE. THIS HELPS IN CONTROLLING THE SPATIAL SIZE OF THE OUTPUT FEATURE MAP. COMMON TYPES INCLUDE:

- VALID PADDING: NO PADDING; THE OUTPUT FEATURE MAP IS SMALLER THAN THE INPUT.
- SAME PADDING: PADDING IS ADDED SO THAT THE OUTPUT FEATURE MAP HAS THE SAME DIMENSIONS AS THE INPUT.

EXAMPLE:

USING A 5X5 IMAGE AND A 3X3 FILTER:

- WITHOUT PADDING & STRIDE 1: OUTPUT IS 3X3.
- WITH SAME PADDING & STRIDE 1: OUTPUT REMAINS 5X5. RESEARCHGATE
- WITH SAME PADDING & STRIDE 2: OUTPUT REDUCES TO 3X3.



CONVOLUTIONAL LAYER



DEEPFLOW

(a) Stride = 1

Diagram illustrating a 1D convolution operation with a 6x6 input grid, a 3x3 kernel, and a 4x4 output grid. The input grid is:

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

The kernel is:

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

The output grid is:

-14	-1	10	-1
-11	-11	7	12
-7	-10	1	13
5	-16	-4	10

(b) Stride = 2

Diagram illustrating a 1D convolution operation with a 6x6 input grid, a 3x3 kernel, and a 2x2 output grid. The input grid is:

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

The kernel is:

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

The output grid is:

-14	10
-7	1

Created by brilliantcode.net



CONVOLUTIONAL LAYER



DEEPFLOW

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

Input Image



0	0	1
1	0	0
0	1	1

Feature
Detector



0	1	0	0	0
0	1	1	1	0
1	0	1	2	1
1	4	2	1	0
0	0	1	2	1

Feature Map

CONVOLUTIONAL LAYER



DEEPFLOW

filters

★
(conv)

1 x1	1 x0	1 x1	0	1
0 x1	1 x0	0 x1	1	1
1 x0	0 x1	1 x0	1	0
1	1	0	0	1
0	1	0	1	0

input



output

CONVOLUTIONAL LAYER



DEEPFLOW

Source layer

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

Convolutional
kernel

-1	0	1
2	1	2
1	-2	0

Destination layer

	5						

$$\begin{aligned} &(-1 \times 5) + (0 \times 2) + (1 \times 6) + \\ &(2 \times 4) + (1 \times 3) + (2 \times 4) + \\ &(1 \times 3) + (-2 \times 9) + (0 \times 2) = 5 \end{aligned}$$

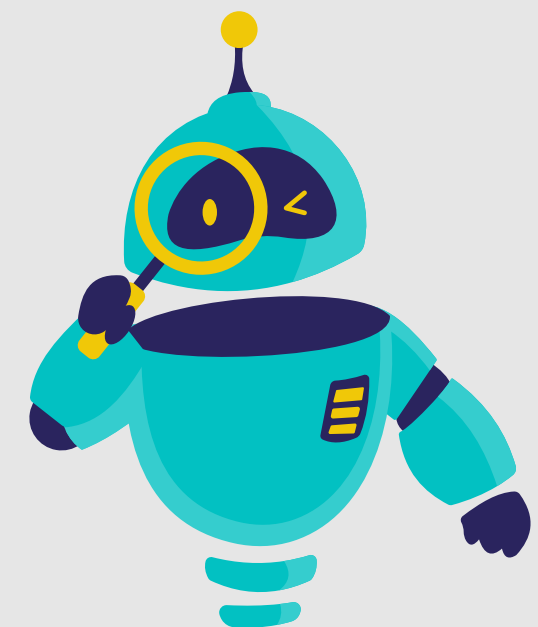
POOLING LAYER



DEEPFLOW

WHAT IS A POOLING LAYER?

A POOLING LAYER REDUCES THE SPATIAL SIZE (HEIGHT AND WIDTH) OF THE FEATURE MAPS WHILE RETAINING THE MOST IMPORTANT INFORMATION.



POOLING LAYER



DEEPFLOW

WHY IS POOLING USED?

TO REDUCE THE NUMBER OF PARAMETERS AND COMPUTATIONS IN THE NETWORK.

HELPS CONTROL OVERFITTING.

MAKES THE MODEL MORE ROBUST TO TRANSLATIONS AND DISTORTIONS.

TYPES OF POOLING:

MAX POOLING: TAKES THE MAXIMUM VALUE FROM A PATCH.

AVERAGE POOLING: TAKES THE AVERAGE OF THE VALUES IN A PATCH.

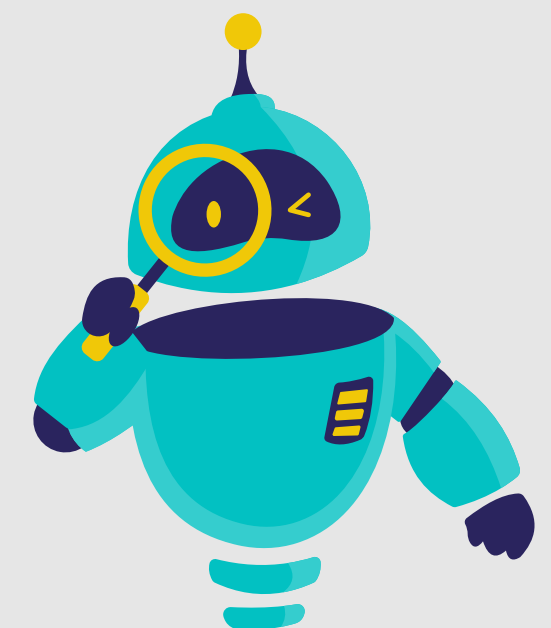
(MAX POOLING IS MOST COMMONLY USED.)

HOW IT WORKS:

A FILTER (E.G., 2X2) SLIDES OVER THE INPUT.

WITH STRIDE (USUALLY 2), IT MOVES STEP-BY-STEP.

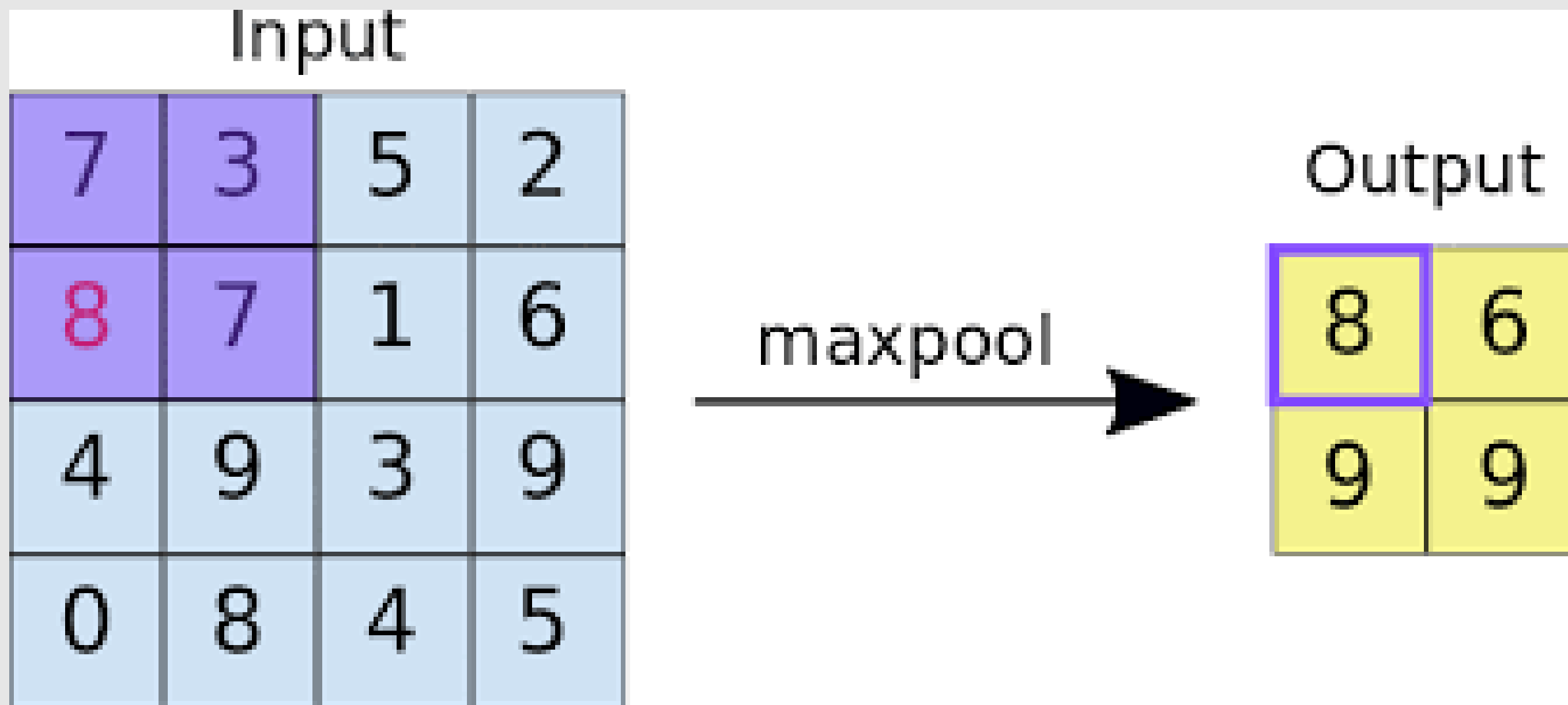
IT OUTPUTS ONE VALUE PER PATCH.



POOLING LAYER



DEEPFLOW



ACTIVATION FUNCTIONS



DEEPFLOW

RELU AND ITS VARIANTS

THE RECTIFIED LINEAR UNIT (RELU) INTRODUCES NON-LINEARITY BY OUTPUTTING ZERO FOR NEGATIVE INPUTS AND THE INPUT ITSELF FOR POSITIVE INPUTS. VARIANTS INCLUDE LEAKY RELU AND PARAMETRIC RELU.

IMPORTANCE OF NON-LINEARITY

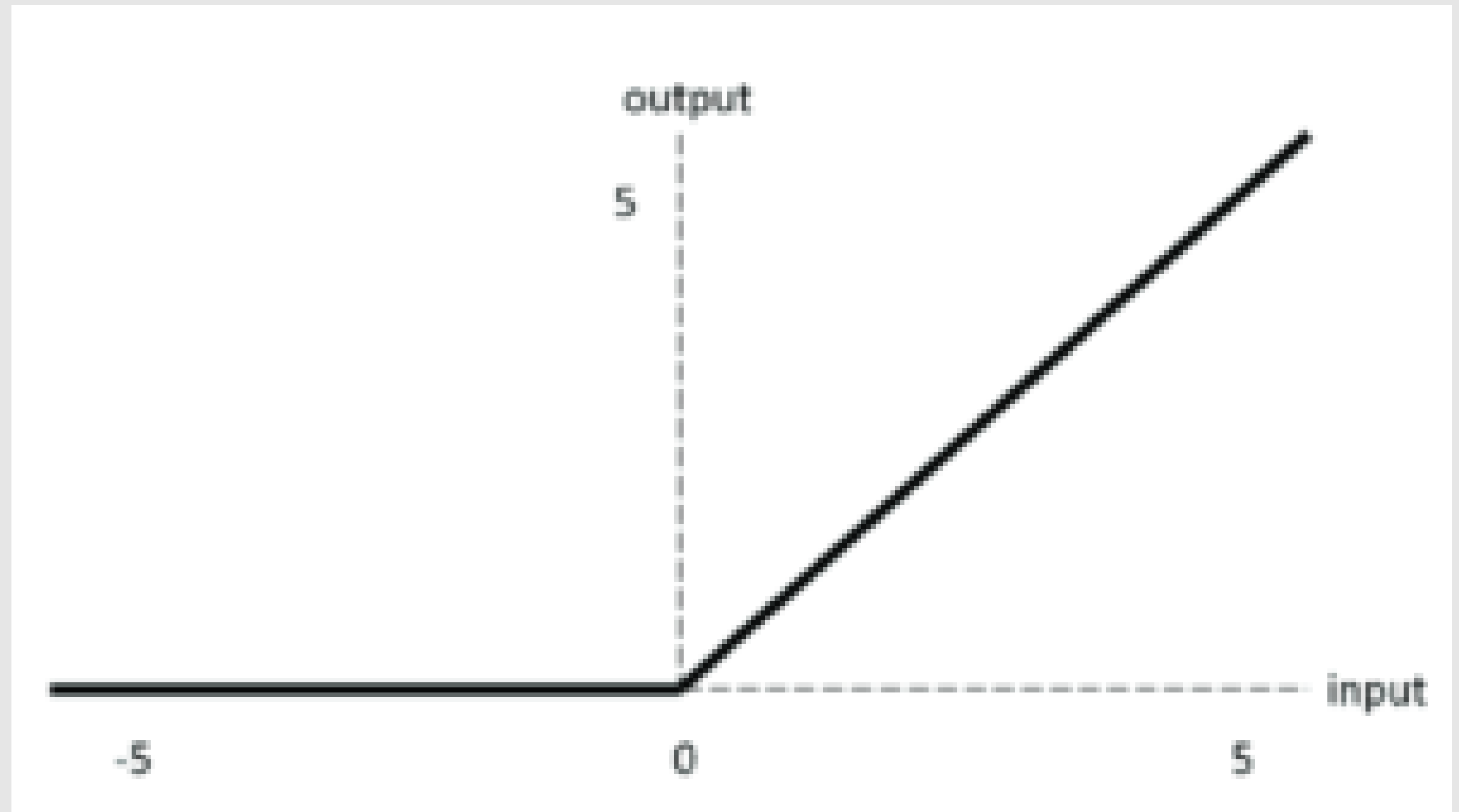
EXPLANATION: NON-LINEAR ACTIVATION FUNCTIONS ENABLE THE NETWORK TO LEARN COMPLEX PATTERNS BEYOND LINEAR RELATIONSHIPS.

ACTIVATION FUNCTIONS



DEEPFLOW

RELU ACTIVATION FUNCTION

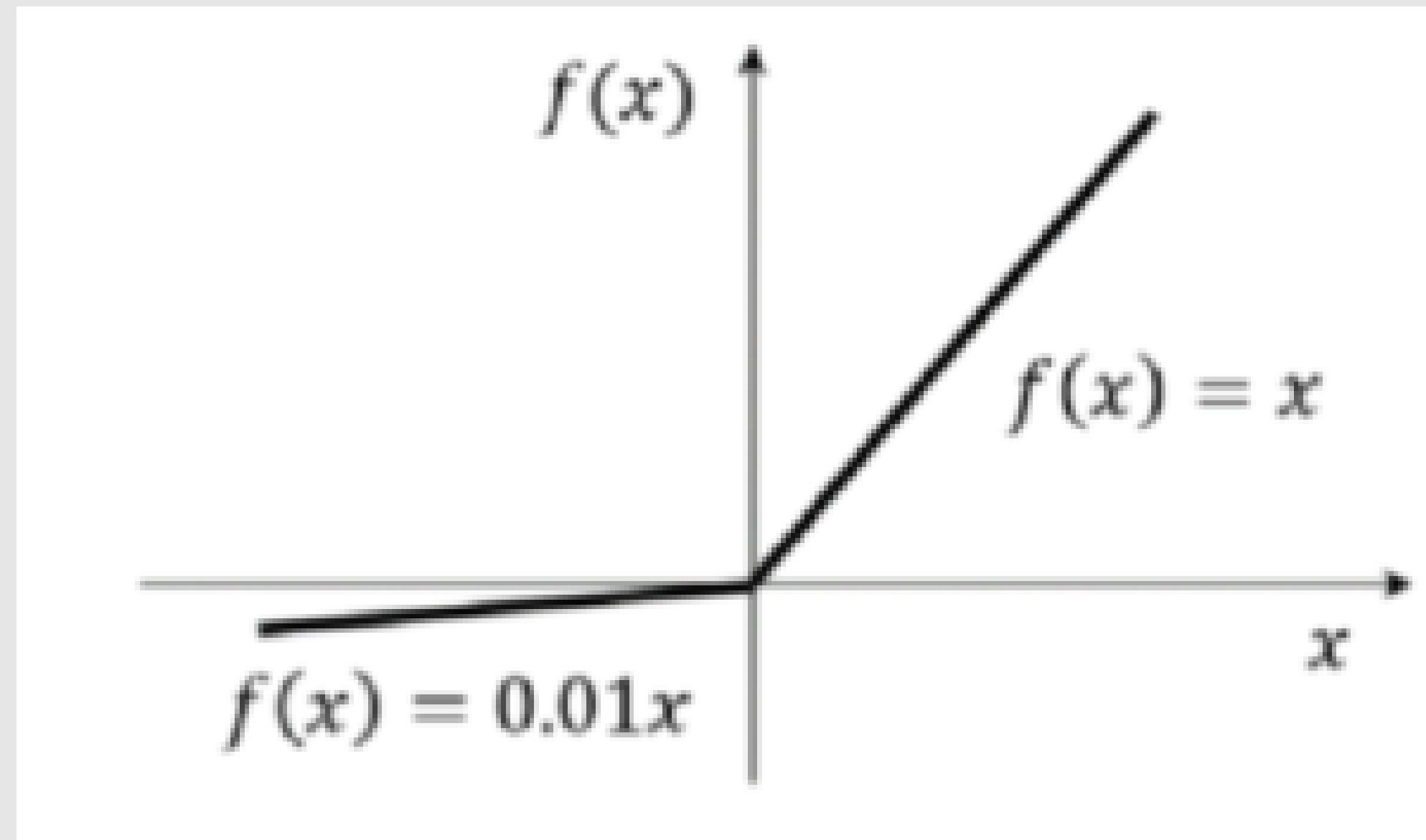


ACTIVATION FUNCTIONS



DEEPFLOW

LEAKY RELU (LRELU)

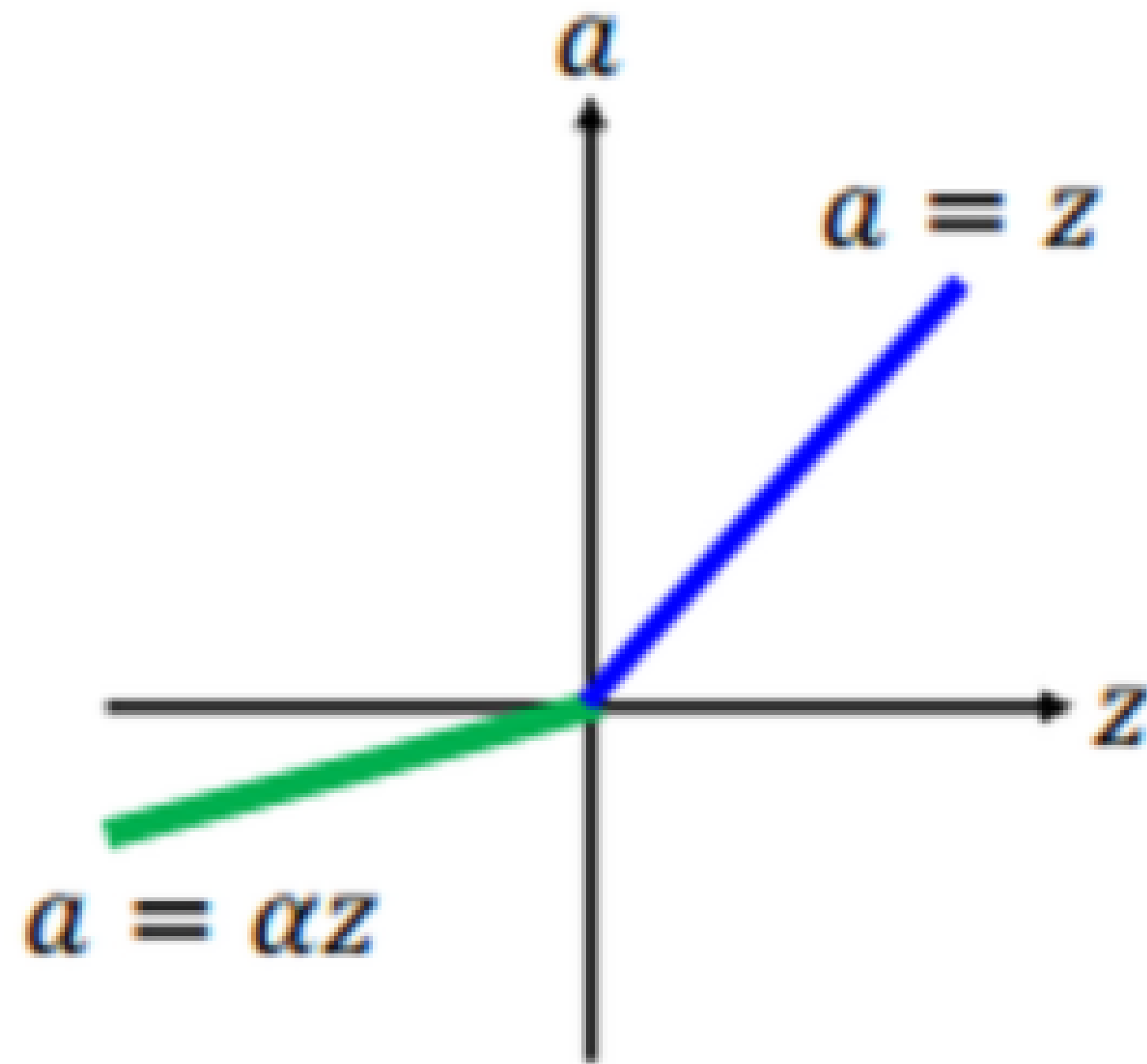


ACTIVATION FUNCTIONS



DEEPFLOW

PARAMETRIC RELU (PRELU)



FLATTENING



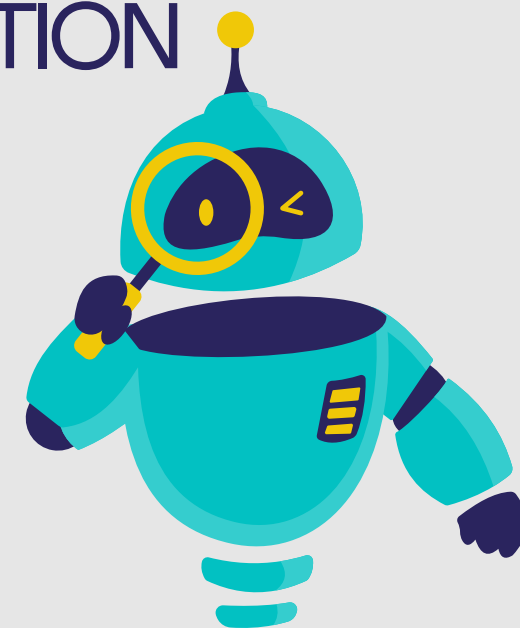
DEEPFLOW

TRANSFORMING FEATURE MAPS TO 1D

CONVERTS THE 2D MATRIX OF FEATURES INTO A 1D VECTOR TO FEED INTO FULLY CONNECTED LAYERS.

WHY IT'S NECESSARY BEFORE FC LAYERS?

FULLY CONNECTED LAYERS REQUIRE A 1D INPUT TO PERFORM CLASSIFICATION BASED ON THE EXTRACTED FEATURES.



FLATTENING

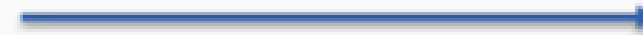


DEEPFLOW

1	1	0
4	2	1
0	2	1

Pooled Feature Map

Flattening



1
1
0
4
2
1
0
2
1

FULLY CONNECTED LAYERS



DEEPFLOW

- **CONNECTING LEARNED FEATURES TO OUTPUT**

THESE LAYERS TAKE THE FLATTENED FEATURE VECTOR AND MAP IT TO THE DESIRED OUTPUT CLASSES.

- **IMPORTANCE OF DENSE LAYERS**

DENSE LAYERS COMBINE FEATURES TO MAKE FINAL PREDICTIONS, ACTING AS THE DECISION-MAKING COMPONENT OF THE NETWORK.

KEY HYPERPARAMETERS IN CNNs



DEEPFLOW

LEARNING RATE

CONTROLS HOW FAST THE MODEL LEARNS. TOO HIGH = UNSTABLE, TOO LOW = SLOW.

EPOCHS

NUMBER OF COMPLETE PASSES THROUGH THE TRAINING DATASET.

BATCH SIZE

NUMBER OF SAMPLES PROCESSED BEFORE MODEL WEIGHTS ARE UPDATED.

KERNEL SIZE

SIZE OF THE CONVOLUTION FILTER (E.G., 3×3, 5×5).

KEY HYPERPARAMETERS IN CNNs



DEEPFLOW

STRIDE

HOW FAR THE FILTER MOVES DURING CONVOLUTION.

PADDING

CONTROLS SPATIAL SIZE OF OUTPUT (E.G., SAME VS VALID).

NUMBER OF FILTERS

DETERMINES HOW MANY FEATURES ARE DETECTED AT EACH LAYER.

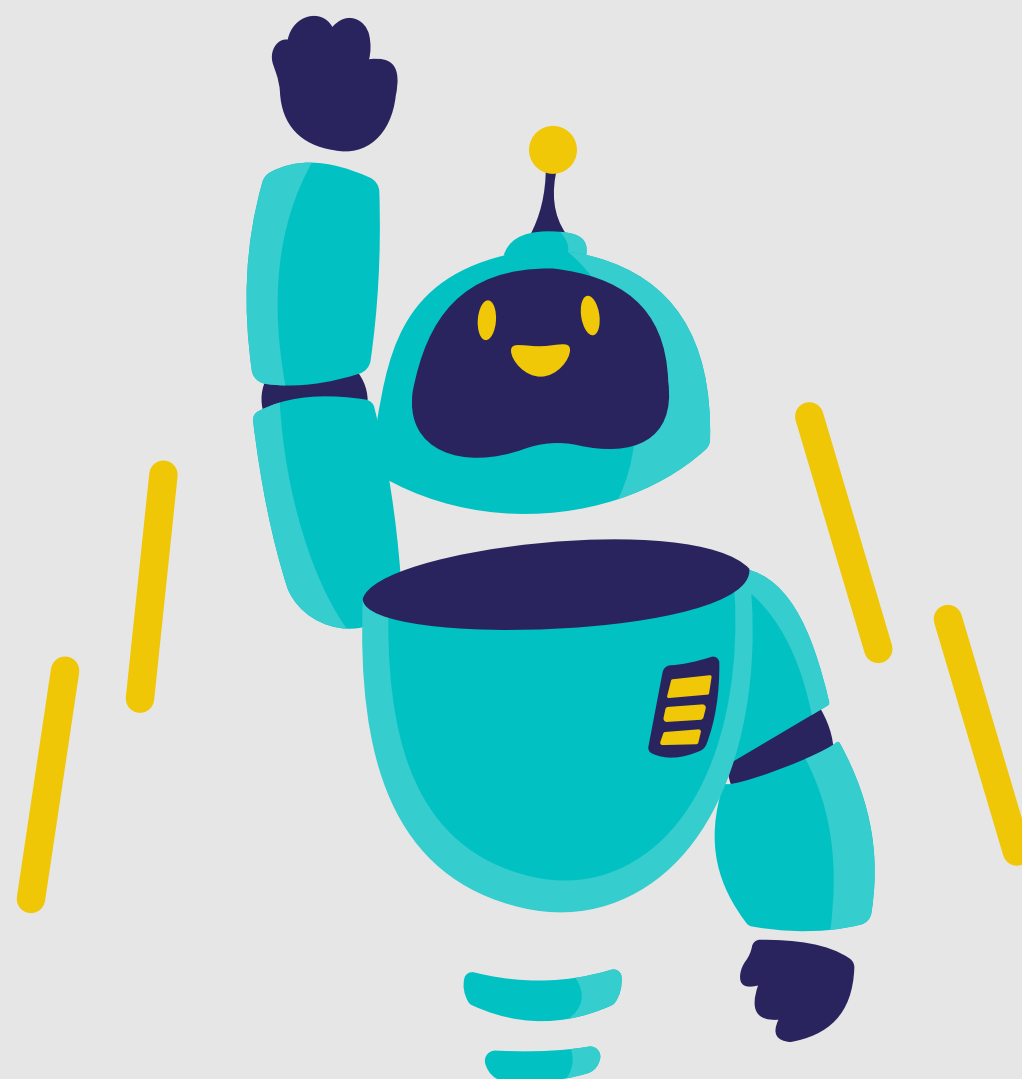
DROPOUT RATE

PREVENTS OVERFITTING BY RANDOMLY DISABLING NEURONS DURING TRAINING.



DEEPFLOW

BASICALLY THAT'S IT 🥰





DEEPFLOW

THANK YOU

SAY “YOUCEP YOU NAILED IT :)”