

A decorative graphic on the left side of the slide consisting of two overlapping parallelograms. The front one is blue and the back one is a light green. They are positioned diagonally, with the blue one partially covering the green one.

Convolutional Neural Network

Computer Vision

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Introduction to Computer Vision

- Computer vision is concerned with the automatic extraction, analysis and understanding of useful information from a single image or a sequence of images.
 - The British Machine Vision Association and Society for Pattern Recognition (BMVA)

(or)

- It is an interdisciplinary field that deals with how computers can be made to gain high-level understanding from digital images or videos.

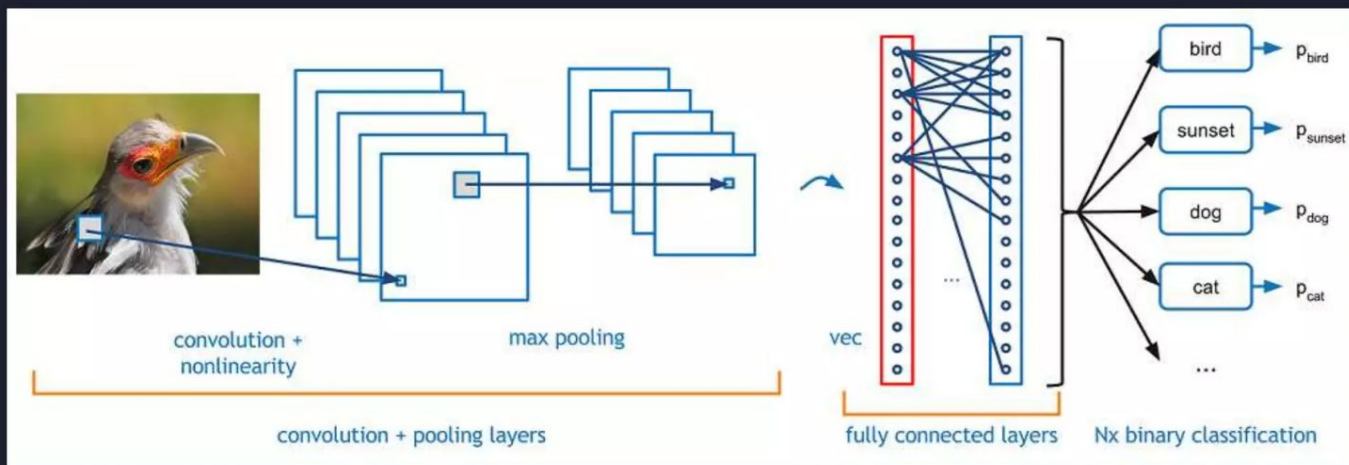
- Wikipedia



What is CNN(Convolution Neural Network)

- It is a class of deep learning.
- Convolutional neural network (ConvNet's or CNNs) is one of the main categories to do images recognition, images classifications, objects detections, recognition faces etc.,
- It is similar to the basic neural network. CNN also have learnable parameter like neural network i.e., weights, biases etc.
- CNN is heavily used in computer vision
- There 3 basic components to define CNN
 - The Convolution Layer
 - The Pooling Layer
 - The Output Layer (or) Fully Connected Layer

Architecture of CNN





Convolution Layer

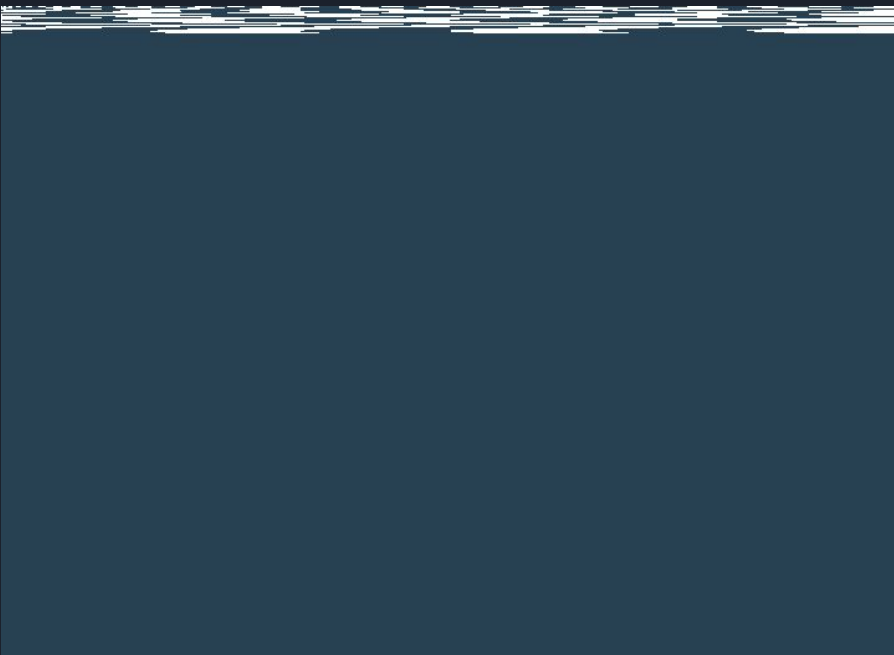
- Computers read images as pixels and it is expressed as matrix ($N \times N \times 3$)—(height by width by depth).
- The Convolutional Layer makes use of a set of learnable filters. A filter is used to detect the presence of specific features or patterns present in the original image (input).
- It is usually expressed as a matrix ($M \times M \times 3$), with a smaller dimension but the same depth as the input file.
- This filter is convolved (slided) across the width and height of the input file, and a dot product is computed to give an activation map.

Convolution Layer

INPUT IMAGE						WEIGHT			
18	54	51	239	244	188	1	0	1	429
55	121	75	78	95	88	0	1	0	
35	24	204	113	109	221	1	0	1	
3	154	104	235	25	130				
15	253	225	159	78	233				
68	85	180	214	245	0				

INPUT IMAGE						WEIGHT			
18	54	51	239	244	188	1	0	1	429
55	121	75	78	95	88	0	1	0	
35	24	204	113	109	221	1	0	1	
3	154	104	235	25	130				
15	253	225	159	78	233				
68	85	180	214	245	0				

Images source: Analytics Vidhya



Small, illegible text or a very thin horizontal strip, possibly a header or a line of code, located at the top of the right half of the image.

Convolution Layer



Original Image



Convolved image

Images source: Analytics Vidhya

Convolution Layer

The concept of stride and padding:

- The weight of a matrix moves 1 pixel at a time is called as stride 1 (as we did in above case).

What if we increase the stride value?

INPUT IMAGE					WEIGHT			
18	54	51	239	244	1	0	1	429
55	121	75	78	95	0	1	0	
35	24	204	113	109	1	0	1	
3	154	104	235	25				
15	253	225	159	78				

Images source: Analytics Vidhya



Convolution Layer

- As we can see in above image the increase in the stride value decreases the size of the image (which may cause in losing the features of the image).
- Padding the input image across it solves our problem, we add more than one layer of zeros around the image in case of higher stride values.

0	0	0	0	0	0	0	0
0	18	54	51	239	244	188	0
0	55	121	75	78	95	88	0
0	35	24	204	113	109	221	0
0	3	154	104	235	25	130	0
0	15	253	225	159	78	233	0
0	68	85	180	214	245	0	0
0	0	0	0	0	0	0	0



Convolution Layer

- when the input of 6x6 is padded around with zeros we get the output with same dimensions of 6x6 this is known as '**Same Padding**'.
- The middle 4x4 pixel remains the same, here we have retained the more information from borders and also preserved the size of image.

0	0	0	0	0	0	0	0
0	18	54	51	239	244	188	0
0	55	121	75	78	95	88	0
0	35	24	204	113	109	221	0
0	3	154	104	235	25	130	0
0	15	253	225	159	78	233	0
0	68	85	180	214	245	0	0
0	0	0	0	0	0	0	0

WEIGHT

1	0	1
0	1	0
1	0	1

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The Pooling Layer

- It can be seen in between the convolution layers in CNN architecture.
- This layer basically reduces the amount of parameters and computation in the network.
- Pooling is done for the sole purpose of reducing the spatial size of the image.
- Pooling is done independently on each depth dimension, therefore the depth of the image remains unchanged. The most common form of pooling layer generally applied is the max pooling.

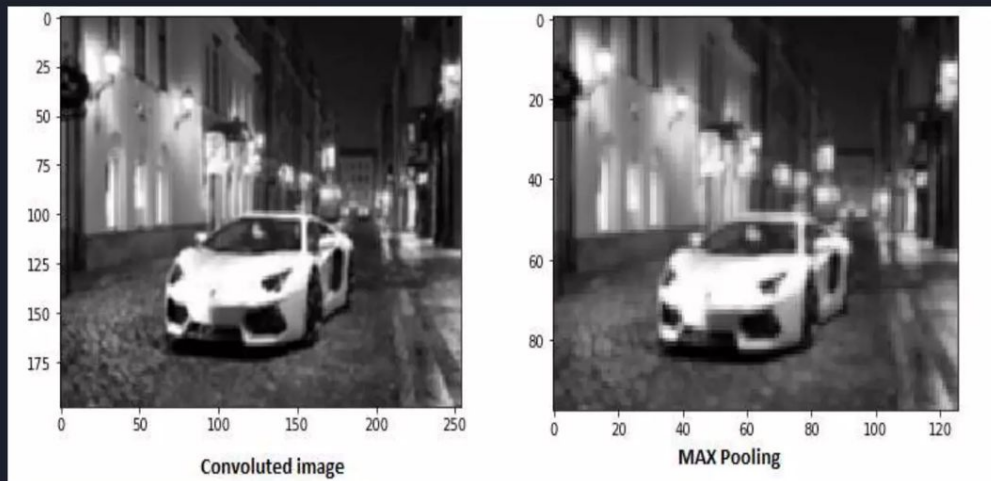
429	505	686	856
261	792	412	640
633	653	851	751
608	913	713	657

792	856
913	851



The Pooling Layer

- Let's try to understand from below image how max pooling has affected our image below.



Images source: Analytics Vidhya



The Output Layer

- After multiple layers of convolution and padding, we would need the output in the form of a class.
- The convolution and pooling layers would only be able to extract features and reduce the number of parameters from the original images.
- However, to generate the final output we need to apply a fully connected layer to generate an output equal to the number of classes we need.
- Convolution layers generate 3D activation maps while we just need the output as whether or not an image belongs to a particular class.
- The output layer will have the loss function like categorical cross-entropy, to compute the error in prediction.
- Once the forward pass is complete the backpropagation begins to update the weight and biases for error and loss reduction.



CNN

How to decide the number of convolution layers and number of filters in CNN ?

- Deeper networks is always better, at the cost of more data and increased complexity of learning.
- You should initially use fewer filters and gradually increase and monitor the error rate to see how it is varying.
- Very small filter sizes will capture very fine details of the image. On the other hand having a bigger filter size will leave out minute details in the image.



Types of CNN

- Based on the problems, we have the different CNN's which are used in computer vision.
- The five major computer vision techniques which can be addressed using CNN.
 - Image Classification
 - Object Detection
 - Object Tracking
 - Semantic Segmentation
 - Instance Segmentation



Applications of Computer Vision

- Computer vision, an AI technology that allows computers to understand and label images, is now used in convenience stores, driverless car testing, daily medical diagnostics, and in monitoring the health of crops and livestock.
- Different use cases found in the computer vision as follows
 - Retail and Retail Security
 - Automotive
 - Healthcare
 - Banking
 - Agriculture
 - Industrial

END

