

→ why FCN Not Good

## Lecture #27 ⇒ CNN (Convolutional Neural Network)

- Deep Learning (Study of Neural Network)
- study the basic neural network (in previous lecture)
- hyperparameters and parameters  
(can not learn from data) (learn during data)
- Calculation phase
  - forward pass → inference (prediction)  
loss calculate
  - backward pass (deep neural network actually learn as base of loss value)

### ⇒ Limitation of Simple Neural Network

(FCN ⇒ fully connected neural)

image is huge array of numbers.

in simple neural network each neuron is connected to its next layers.

increasing each neuron mean that is connected with all other neurons in its next layer.

so it will be very huge even if image is very low

◦ Simple neural network treat all pixels in an image equally.

◦ Can not effectively capture the local pattern and structures in images

Fully connected connection :- number of weights increased <sup>drastically</sup>



Features  $\Rightarrow$  images are full of important part called feature.  
it can be edge, texture, shapes or even face.

$\hookrightarrow$  neuron perform convolutional operations.  $\hookrightarrow$  To perform convolution we need filter.

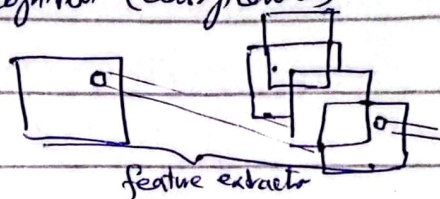
$\Rightarrow$  CNN (for images & videos)

$\therefore$  for tasks like image recognition (classification)

### Basic structure

#### Convolutional layer

detect features like edges, texture & patterns in an image.



#### Convolutional operation

- was using in image processing
- so this operation used

" apply filters to the input image to create feature maps.

filters are small matrices that slides over the input image & detect patterns.  
usually ReLU (Rectified Linear Unit) that introduces non-linearity.

• we apply filter on image

• example vertical edge detector

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



- instead of convolving image into single array we are now (in FCN)
- picture is its in structure & we apply filter on it (matrix-matrix multiplication)
- learning filter (CNN) finding weight (FCN)

max neuron is a convolutional neuron which

وہ جو رزلٹ آگے گا فلاں کا وہ اگلی سیر کے دوران  
 ہو جائے وہ اور فلاں ایلائی رنگی —

(Feature Map) ← image کے ادھر فلٹر اپڈیٹ کرنے کے بعد

Map of particle features

Effect: when apply on image  $\Rightarrow$  map of all vertices bones

$3 \times 3$ 

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



## what we learn

initial goal → input image, feature extract

↓  
filters applied

↓  
gives feature maps (a output)

(that contain features of images)

یعنی ہر لیئر (feature) نکالتی رہتی

ہے اور آگے بھیجتی رہتی ہے۔۔۔ اس طرح

کرتے آگے میں ہمارے پاس

(pooling) و ۰.۵ feature (اس کی summarizing)

کے آگے پاس کرتی ہے

input convolution pooling  
↓  
feature extractor

۱. پہلی لیئر میں (simple feature) نکالتے

۲. اسے آگے لیئر میں ان کے ملکر اور تھوڑے (complex)

Feature نکالتے

۳. اس سے آگے لیئر میں: پہلی feature کے دیکھ کر مزید (complex) نکالتے

۴. اور آخر میں اس کی Base پر ہم فائنل رزلٹ دیکھ لیتے ہیں

## Summary

### Convolutional Neural Network

- important parts
  - (i) convolutional layer (Take image as input, simple apply convolution and extract features)
  - (ii) Pooling layer (Take feature as input and forward summarized to next feature map)
  - (iii) Flatten layer (last pooling layer to input of 3D converted)
  - (iv) Read → Neural Network (A fully connected neural network)  
۳. ۵۰ (flatten layer) کے ۱۲۰, ۶۰, ۲۰