12.12.23

CSE-461

"Lecture-13"

"Intro to CNNs and Object Detection"

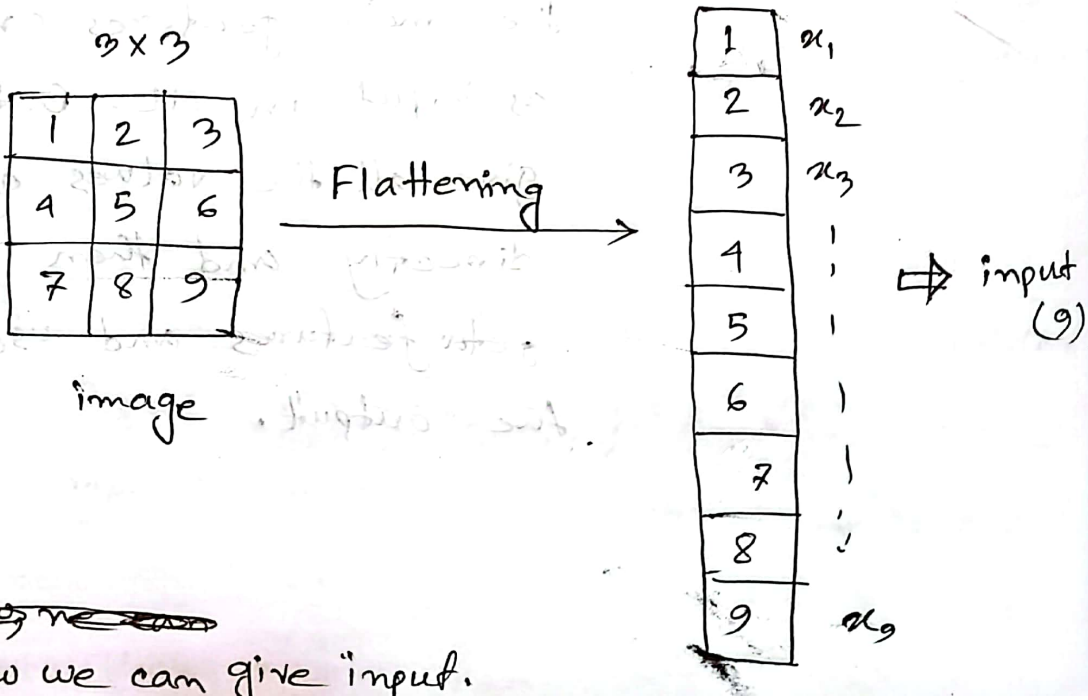
→ Input features are two types,

(a) Features are given (uses ML)

(b) Raw data is given not features (uses DL)

⇒ Now, how to give input? "Pixels as Features"

(a) we can do Flattening of our image.



now we can give input.

→ But we have a problem. The problem is ~~time~~ dimensionality.

if we have 3×3 image, then input = 9

" " " 12×12 " " " = 144

⋮ ⋮

" " " 680×420 " " " = 856,800

→ So, image size বাড়লে input features বাড়তেছে and for that ~~we have~~ the cost is increasing and also time consuming.

→ Second is, একটা image থেকে আমরা actually কোনো feature পাবো না। একটা image থেকে object ~~detect~~ করা কঠিন। So, image flatten করলে structural information থাকতেছে না।

→ So, to Solve these two Problems we will use 'Convolution'.

→ **Convolution:**

Here, we have a image and a filter or Kernel. So, we will place the kernel in our main image in the upper left corner. Then the image and the Kernel এর pixel value গুলো multiply করে sum করবো and সেই value পাবো সেটা feature Matrix এর upper left pixel এ বসাবে।

Then kernel কে 1 ঘর right এ বসাবো and then same ভাবে করে একটা feature Matrix পাবো।

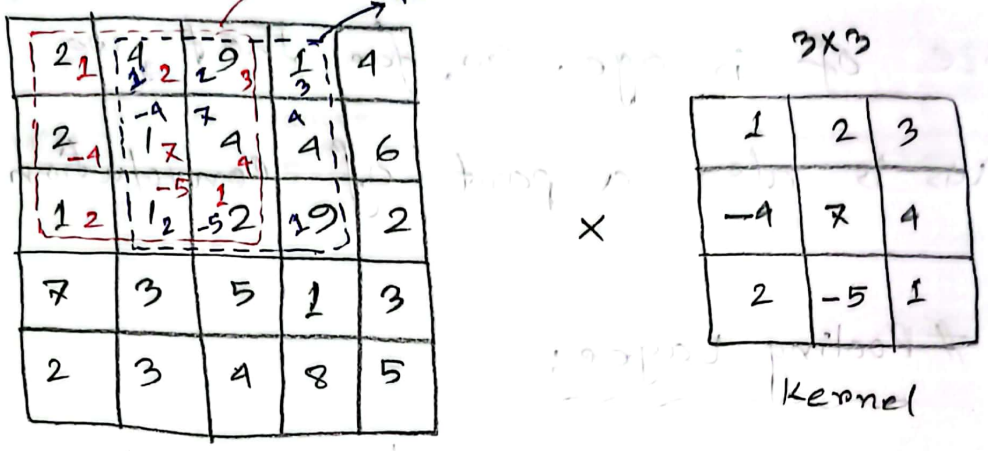
→ So, feature Matrix এর dimension কি হবে?

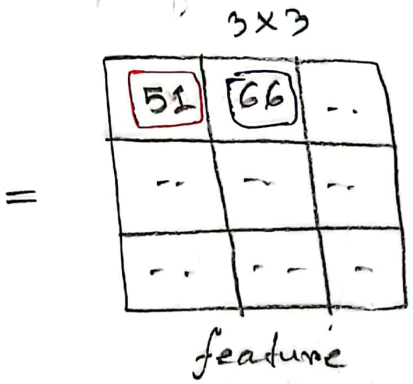
⇒ (image dimension − kernel dim) + 1 = FM

∴ feature dim = $(n - k + 1)$

multiply pixels and the sum = 51

multiply pixels and the sum = 66





here,

$n = 5$

$k = 3$

$$\therefore \text{feature} = (n - k + 1)$$
$$= (5 - 3 + 1)$$
$$= 3$$

So, convolution করে আমরা feature কে small pixels এ convert করে দিচ্ছি and যার কারণে flatten করলে input feature বড় করা যাবে as input.

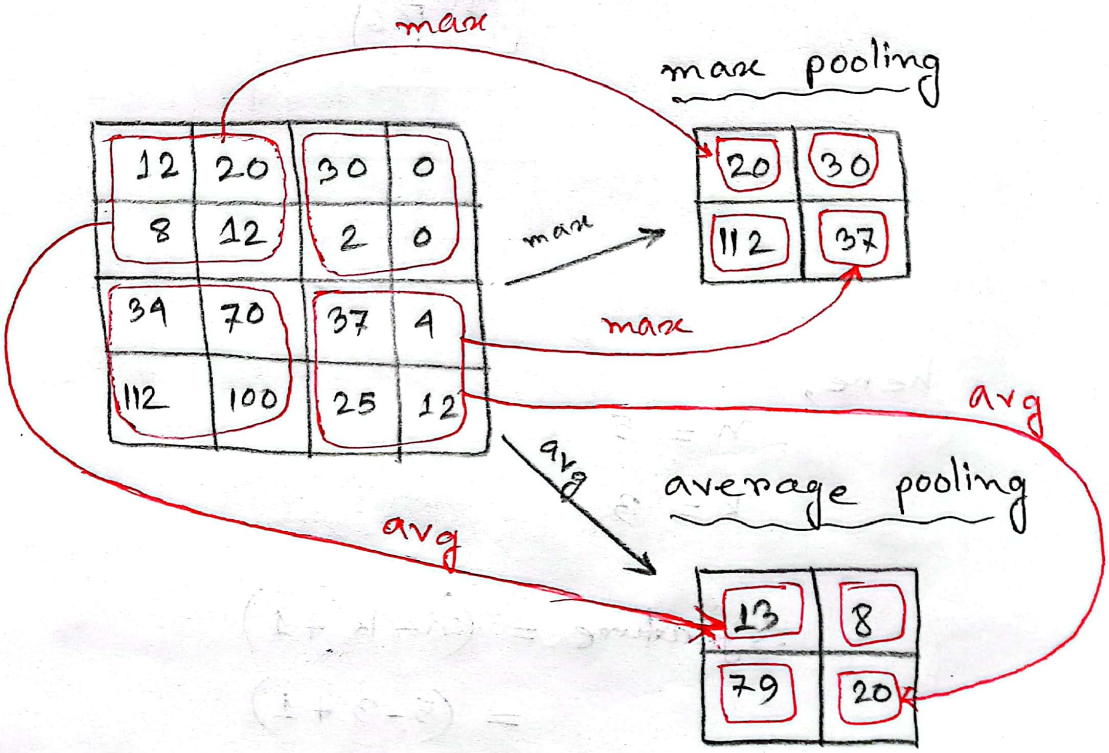
CNN will learn which Kernel should use.

Again, in practical life we will have a large size of image. So, for that we use Pooling. This is also a part of Convolution.

Pooling Layer:

(a) max pooling

(b) average pooling



→ By doing pooling we also reduce the number of parameters in case of large Image. Also, retains major information.

→ Architecture of CNNs: