



Type something...

01) Convolution

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Convolution?

이미지 위에서 stride 값 만큼 filter(kernel)을 이동시키면서
겹쳐지는 부분의 각 원소의 값을 곱해서 모두 더한 값을 출력으로
하는 연산

1	2	3	1	0	1
0	1	5	0	1	0
1	0	2	2	1	0

8

1	2	3	0	1
0	1	5	1	0
1	0	2	2	1
1	1	2	0	0
1	0	1	1	1

input

1	0	1
0	1	0
1	0	1

filter

8	9	8
8	5	9
6	5	5

output

$$(1 \times 1) + (2 \times 0) + (3 \times 1) + \\ (0 \times 0) + (1 \times 1) + (5 \times 0) + \\ (1 \times 1) + (0 \times 0) + (2 \times 1) = 8$$

- Convolution은 filter를 stride 값만큼 이동시켜 각 원소의 값을 곱해서 모두 더한 값을 만드는 연산이다.

Stride and Padding

- stride : filter를 한번에 얼마나 이동 할 것인가
- padding : zero-padding

0	0	0	0	0	0	0
0	1	2	3	0	1	0
0	0	1	5	1	0	0
0	1	0	2	2	1	0
0	1	1	2	0	0	0
0	1	0	1	1	1	0
0	0	0	0	0	0	0

input + padding

- stride: filter를 얼마나 이동시킬 것인가, padding : 테두리에 zero를 얼마나 넣을 것인가(이미지 손실을 방지하기 위함임)

Conv2d ⚡



```
CLASS torch.nn.Conv2d(in_channels, out_channels, kernel_size, stride=1, padding=0, dilation=1,  
groups=1, bias=True, padding_mode=‘zeros’, device=None, dtype=None) [SOURCE]
```

Applies a 2D convolution over an input signal composed of several input planes.

In the simplest case, the output value of the layer with input size (N, C_{in}, H, W) and output ($N, C_{\text{out}}, H_{\text{out}}, W_{\text{out}}$) can be precisely described as:

$$\text{out}(N_i, C_{\text{out}_j}) = \text{bias}(C_{\text{out}_j}) + \sum_{k=0}^{C_{\text{in}}-1} \text{weight}(C_{\text{out}_j}, k) * \text{input}(N_i, k)$$

- `torch.nn.Conv2d(in_channels, out_channels, kernel_size, stride, padding,)`

입력의 형태

- input type : `torch.Tensor`
- input shape : ($N \times C \times H \times W$)
(batch_size, channel, height, width)

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- input shape = ($N \times C \times H \times W$): (batch_size, channel, height, width)

Convolution의 output 크기

$$\text{Output size} = \frac{\text{input size} - \text{filter size} + (2 * \text{padding})}{\text{Stride}} + 1$$

예제 1)
input image size : 227 x 227
filter size = 11x11
stride = 4
padding = 0
output image size = ?

예제 2)
input image size : 64 x 64
filter size = 7x7
stride = 2
padding = 0
output image size = ?

예제 3)
input image size : 32 x 32
filter size = 5x5
stride = 1
padding = 2
output image size = ?

예제 4)
input image size : 32 x 64
filter size = 5x5
stride = 1
padding = 0
output image size = ?

예제 5)
input image size : 64 x 32
filter size = 3x3
stride = 1
padding = 1
output image size = ?

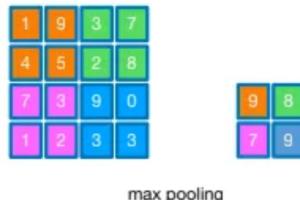
$$\text{output size} = \frac{\text{input size} - \text{filter size} + 2 * (\text{padding})}{\text{Stride}} + 1$$

- 만약 나눗셈 과정에서 28.5 이런식으로 남았으면 버려주고 28만 사용한다. (내림)

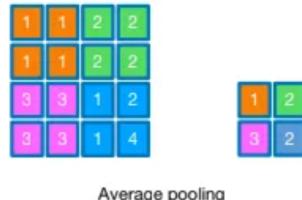


Pooling

- Max Pooling

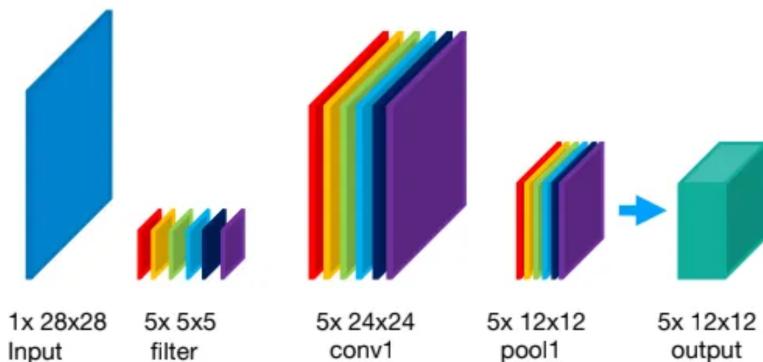


- Average Pooling



- pooling은 구역별로 값을 추출하는 과정으로 max pooling은 최대값을 추출, average pooling은 평균값을 추출한다.

CNN implementation



```
input = torch.Tensor(1,1,28,28)  
conv1=nn.Conv2d(1,5,5)  
pool = nn.MaxPool2d(2)  
out = conv1(input)  
out2 =pool(out)  
out.size()  
out2.size()
```

- batch_size를 제외하고 channel * height * stride만 고려한 필터이다.
- Input: 1 x 28 x 28
- filter: 5 x 5 x 5 `nn.Conv2d(1, 5, 5)` :`input_channel = 1, output_channel = 5, kernel_size = 5`
- conv1: 5 x 24 x 24
- `pool1 = nn.MaxPool2d(2)` :`5x12x12 output`

