

# **Deep Learning**

4.2 Pooling

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- Groups multiple activations and replaces by a representative one
- 2 Reduces the dimensionality of the signal progressively  $\rightarrow$  considers non-overlapping stride
- ③ Also called sub-sampling layer
- ④ Generally found between two convolution layers (and parameter free)

#### Max Pooling



Standard in CNNs

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Standard in CNNs

2 Computes maximum value over a non-overlapping blocks in the input

			1	Input (widt	hW)		
1	2	-1	0	2	3	-2	0

	Outp	ut (width V	V/w)	~
2	0	3	0	

#### **Average Pooling**



#### Computes the average of the receptive field

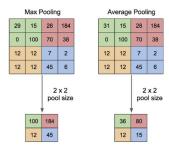
Input (width W)							
1	2	-1	0	2	3	-2	0

	Outpu	t (width W	/w)	
1.5	-0.5	2.5	-1	

#### Pooling in 2D



#### Isame as 1D, but the receptive field is 2D and non-overlapping



#### Figure credits: Preston Hoang and Quora

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#### Pooling in 2D

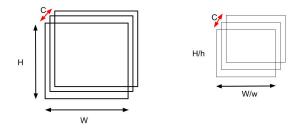


Contrary to Convolution, Pooling applies channel wise

#### Pooling in 2D



- Ontrary to Convolution, Pooling applies channel wise
- 2 No reduction in number of channels, only spatial size reduction



#### Pooling provides weak invariance



1 Operation is invariant to any permutation within the block

#### Pooling provides weak invariance



- Operation is invariant to any permutation within the block
- 2 Withstands deformations caused by local translations



F.max\_pool2d(input, kernel\_size, stride=None, padding=0, dilation=1, ceil\_mode=False, return\_indices=False)

Applies max pooling on each of the channels separately



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- Applies max pooling on each of the channels separately
- 2 input is  $N \times C \times H \times W$  tensor
- $\ensuremath{\mathfrak{G}}$  kernel\_size is (h,w) or k
- ④ Result would be a tensor of size  $N \times C \times \lfloor H/h \rfloor \times \lfloor W/w \rfloor$

#### Pooling in PyTorch



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- Default stride is the kernel size (for convolution, it is 1)
- ② But, it can be modulated if required
- ③ Default padding is zero

#### Pooling Layer in PyTorch



class torch.nn.MaxPool2d(kernel\_size, stride=None, padding=0, dilation=1, return\_indices=False, ceil\_mode=False)