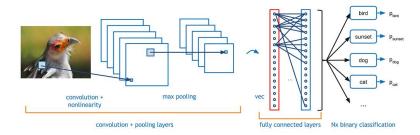


## **Deep Learning**

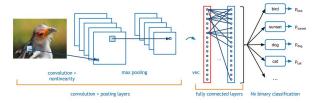
4.3 Putting it all together - CNN

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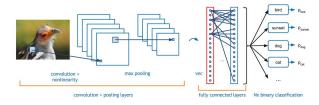






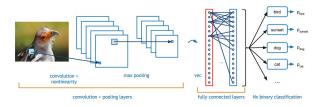
Initially Conv layer with nonlinearity





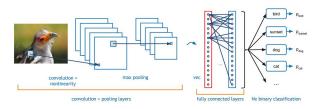
- Initially Conv layer with nonlinearity
- 2 Followed by a few Conv + Nonlinearity layers





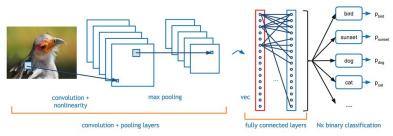
- Initially Conv layer with nonlinearity
- Followed by a few Conv + Nonlinearity layers
- $\ensuremath{\mathfrak{G}}$  Have Pooling layers in between Conv layers  $\to$  reduce the feature map size sufficiently



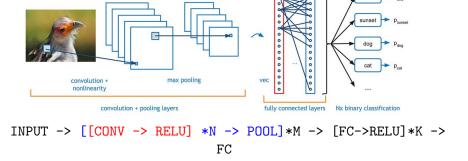


- Initially Conv layer with nonlinearity
- 2 Followed by a few Conv + Nonlinearity layers
- 4 Vectorize and and fully connected layers

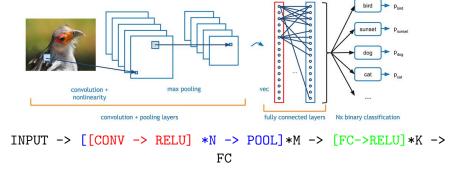












input size/ layer information	output size	# parameters	# products
$1 \times 28 \times 28$			
nn.Conv2d(1, 32, kernel_size=5)			

input size/ layer information	output size	# parameters	# products
$1 \times 28 \times 28$	$32 \times 24 \times 24$		
nn.Conv2d(1, 32, kernel_size=5)			

input size/ layer information	output size	# parameters	# products
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2+1)=832$	
nn.Conv2d(1, 32, kernel_size=5)			

input size/ layer information	output size	# parameters	# products
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2+1)=832$	$32.24^2.5^2 = 460800$
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			
F may nool2d( kernel size=3)			

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input size/ layer information	output size	# parameters	# products
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2+1)=832$	$32.24^2.5^2 = 460800$
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			
E mar noolod( kornol giro-3)	32 ~ 8 ~ 8	0	0

input size/ layer information	output size	# parameters	# products
$1\times28\times28\\ \texttt{nn.Conv2d(1, 32, kernel\_size=5)}$	$32 \times 24 \times 24$	$32.(5^2 + 1) = 832$	$32.24^{2}.5^{2} = 460800$
$32 \times 24 \times 24$			
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$ / F.relu(.)	$32 \times 8 \times 8$	0	0

input size/ layer information	output size	# parameters	# products
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2 + 1) = 832$	$ 32.24^{2}.5^{2} \\ = 460800 $
<pre>nn.Conv2d(1, 32, kernel_size=5)</pre>			
$32 \times 24 \times 24$			
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8 / \text{F.relu(.)}$	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$			
nn.conv2d(32, 64, kernel size=5)	$64 \times 4 \times 4$	$64.(32.4^2 + 1) = 51264$	$64.32.4^{2}.5^{2} = 819200$

input size/ layer information	output size	# parameters	# products
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2 + 1) = 832$	$32.24^2.5^2$ = 460800
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8 / \text{F.relu(.)}$	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$			
		$64.(32.4^2+1)$	64.32.4 <sup>2</sup> .5 <sup>2</sup>
nn.conv2d(32, 64, kernel_size=5)	$64 \times 4 \times 4$	=51264	= 819200
$64 \times 4 \times 4$			
<pre>F.max_pool2d(., kernel_size=2)</pre>	$64 \times 2 \times 2$	0	0

input size/ layer information	output size	# parameters	# products
		(-2 .)	$32.24^2.5^2$
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2+1)=832$	= 460800
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			_
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8 / \text{F.relu(.)}$	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$			
		$64.(32.4^2+1)$	64.32.4 <sup>2</sup> .5 <sup>2</sup>
nn.conv2d(32, 64, kernel_size=5)	$64 \times 4 \times 4$	=51264	= 819200
$64 \times 4 \times 4$			
<pre>F.max_pool2d(., kernel_size=2)</pre>	$64 \times 2 \times 2$	0	0
$64 \times 2 \times 2$ / F.relu(.)	$64 \times 2 \times 2$	0	0

input size/ layer information	output size	# parameters	# products
mput size/ layer information	Output Size	# parameters	32.24 <sup>2</sup> .5 <sup>2</sup>
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2 + 1) = 832$	= 460800
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8 / \text{F.relu(.)}$	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$			
		$64.(32.4^2+1)$	64.32.42.52
nn.conv2d(32, 64, kernel_size=5)	$64 \times 4 \times 4$	=51264	= 819200
$64 \times 4 \times 4$			
<pre>F.max_pool2d(., kernel_size=2)</pre>	$64 \times 2 \times 2$	0	0
$64 \times 2 \times 2$ / F.relu(.)	$64 \times 2 \times 2$	0	0
$64 \times 2 \times 2$	256	0	0
x.view(-1,256)			

input size/ layer information	output size	# parameters	# products
			32.24 <sup>2</sup> .5 <sup>2</sup>
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2 + 1) = 832$	= 460800
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8 / \text{F.relu(.)}$	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$			
		$64.(32.4^2+1)$	64.32.42.52
nn.conv2d(32, 64, kernel_size=5)	$64 \times 4 \times 4$	=51264	= 819200
$64 \times 4 \times 4$			
<pre>F.max_pool2d(., kernel_size=2)</pre>	$64 \times 2 \times 2$	0	0
${64 \times 2 \times 2 \text{ / F.relu(.)}}$	$64 \times 2 \times 2$	0	0
$64 \times 2 \times 2$	256	0	0
x.view(-1,256)			
256	0	0	0
nn.Linear(256,200)	200	200(256+1)=51400	200.256=51200

input size/ layer information	output size	# parameters	# products
			32.24 <sup>2</sup> .5 <sup>2</sup>
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2+1)=832$	= 460800
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8 / \text{F.relu(.)}$	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$			
		$64.(32.4^2+1)$	64.32.4 <sup>2</sup> .5 <sup>2</sup>
nn.conv2d(32, 64, kernel_size=5)	$64 \times 4 \times 4$	=51264	= 819200
$64 \times 4 \times 4$			
<pre>F.max_pool2d(., kernel_size=2)</pre>	$64 \times 2 \times 2$	0	0
${64 \times 2 \times 2 \text{ / F.relu(.)}}$	$64 \times 2 \times 2$	0	0
$64 \times 2 \times 2$	256	0	0
x.view(-1,256)			
256	0	0	0
nn.Linear(256,200)	200	200(256+1)=51400	200.256=51200
200 / F.relu(.)	200	0	0

input size/ layer information	output size	# parameters	# products
·			32.24 <sup>2</sup> .5 <sup>2</sup>
$1 \times 28 \times 28$	$32 \times 24 \times 24$	$32.(5^2+1)=832$	= 460800
nn.Conv2d(1, 32, kernel_size=5)			
$32 \times 24 \times 24$			
<pre>F.max_pool2d(., kernel_size=3)</pre>	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8 / \text{F.relu(.)}$	$32 \times 8 \times 8$	0	0
$32 \times 8 \times 8$			
		$64.(32.5^2+1)$	64.32.42.52
nn.conv2d(32, 64, kernel_size=5)	$64 \times 4 \times 4$	=51264	= 819200
$64 \times 4 \times 4$			
<pre>F.max_pool2d(., kernel_size=2)</pre>	$64 \times 2 \times 2$	0	0
$64 \times 2 \times 2 / \text{F.relu(.)}$	$64 \times 2 \times 2$	0	0
$64 \times 2 \times 2$	256	0	0
x.view(-1,256)			
256	0	0	0
nn.Linear(256,200)	200	200(256+1)=51400	200.256=51200
200 / F.relu(.)	200	0	0
200	0	0	0
nn.Linear(200,10)	10	10(200+1)=2010	10.200=2000

### Recent architectures are more sophisticated



Note that LeNet is a classical architecture and does not reflect the recent CNNs in complexity

#### Recent architectures are more sophisticated



- Note that LeNet is a classical architecture and does not reflect the recent CNNs in complexity
- Recent CNN architectures are far more sophisticated
  - More depth
  - Regularizers to handle the depth