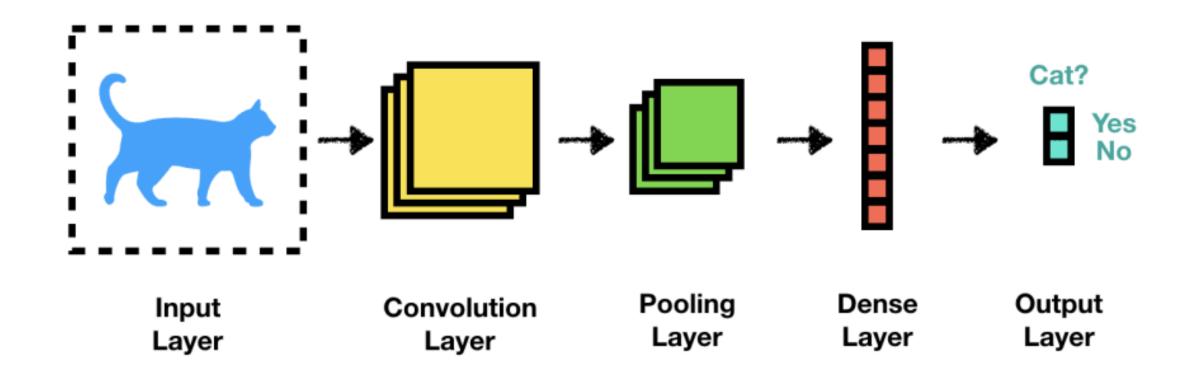
Keras와 함께 하는 딥러닝기초

이정근

소프트웨어 융합 대학 (School of Software)

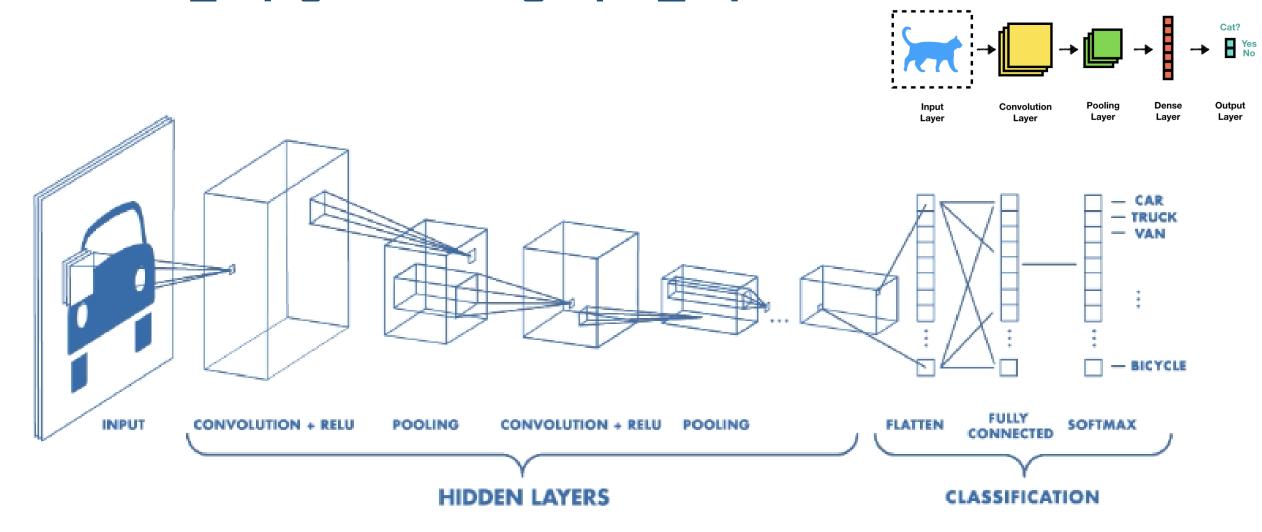
JeongGun_Lee@hallym_ac_kr / www_onchip_net

Convolutional Neural Networks (CNN)



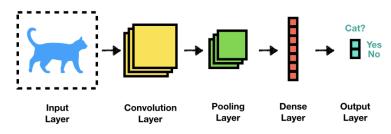
https://github.com/jeonggunlee/DeepLearningBasics/blob/master/Lab07_introduction_to_convnets.ipynb

* https://github.com/rickiepark/deep-learning-with-python-notebooks



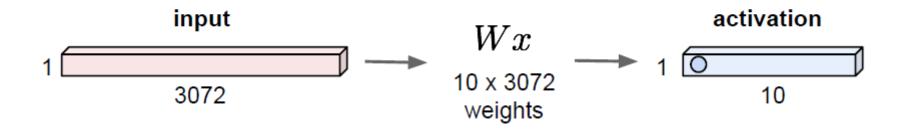
https://github.com/jeonggunlee/DeepLearningBasics/blob/master/Lab07 introduction to convnets.ipynb

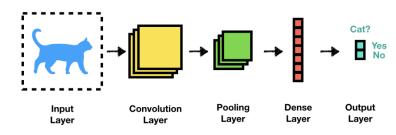
* https://github.com/rickiepark/deep-learning-with-python-notebooks



Fully Connected Layer

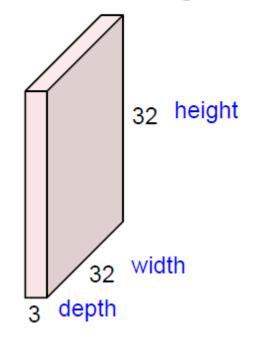
32x32x3 image -> stretch to 3072 x 1

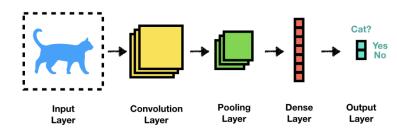




Convolution Layer

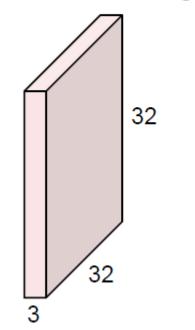
32x32x3 image -> preserve spatial structure



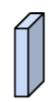


Convolution Layer

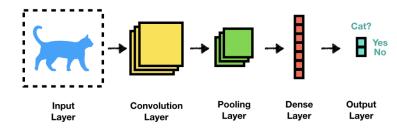
32x32x3 image

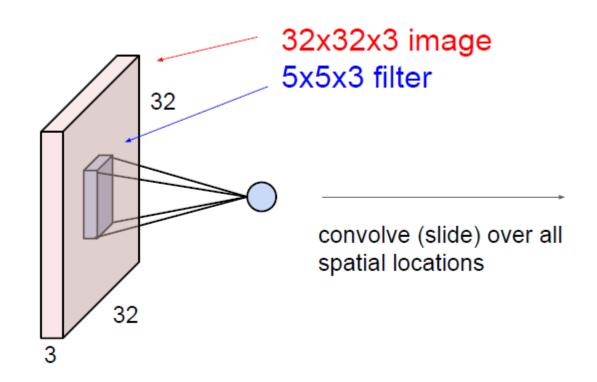


5x5x3 filter

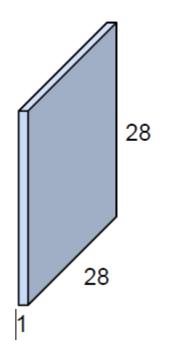


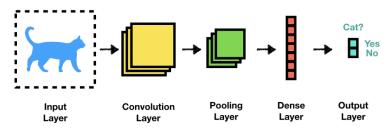
Convolve the filter with the image i.e. "slide over the image spatially, computing dot products"



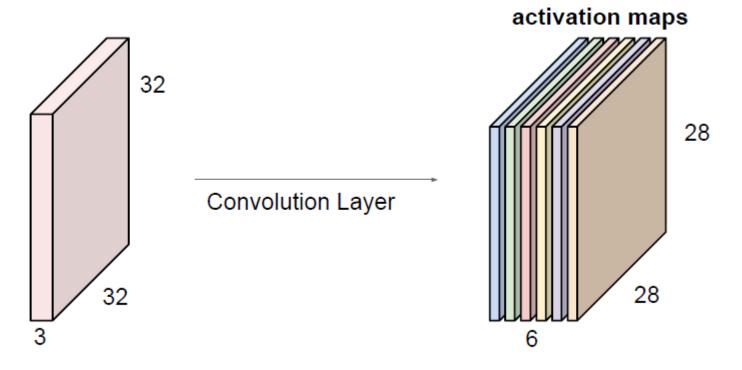


activation map

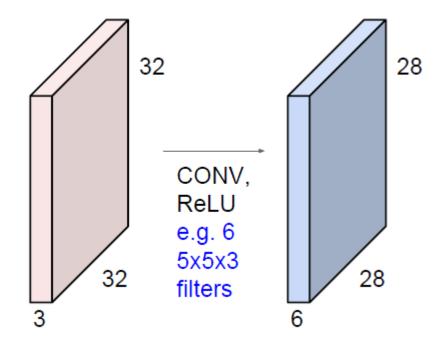


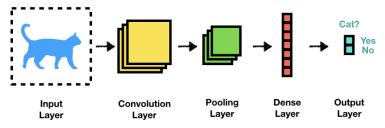


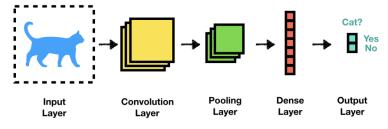
For example, if we had 6 5x5 filters, we'll get 6 separate activation maps:

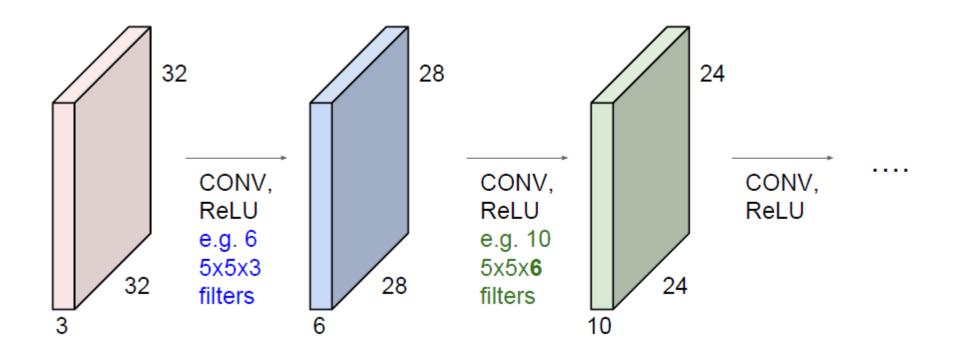


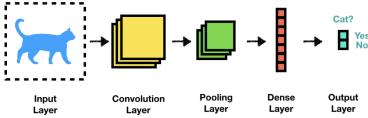
We stack these up to get a "new image" of size 28x28x6!

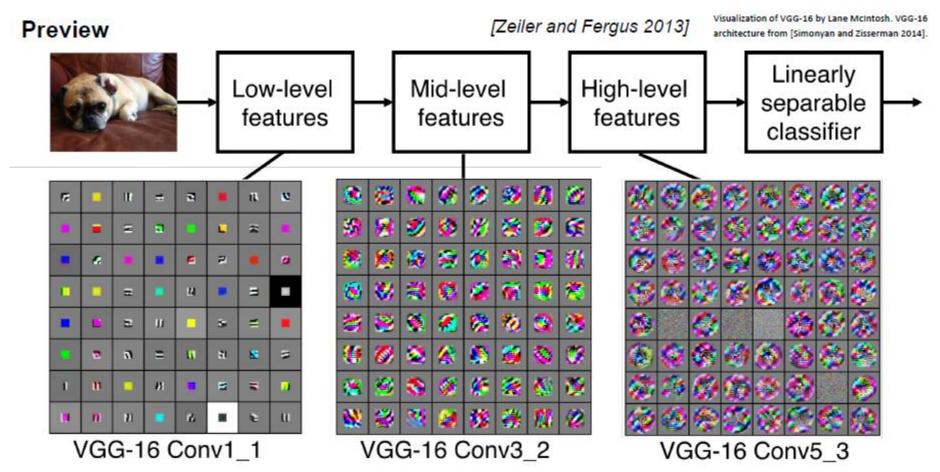


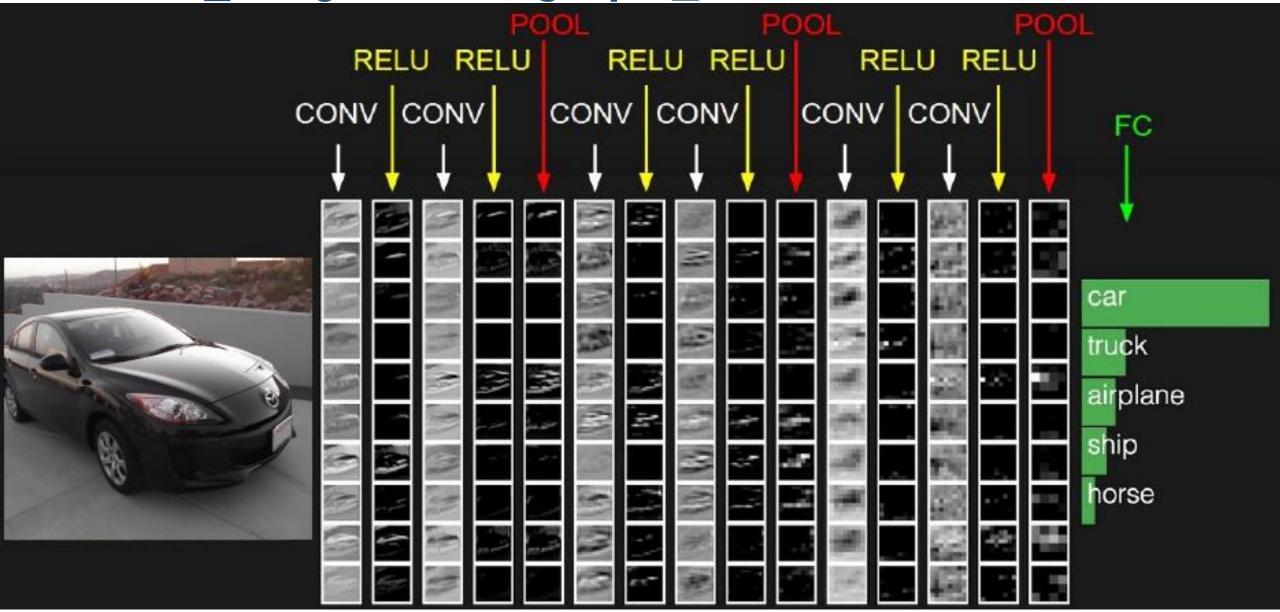




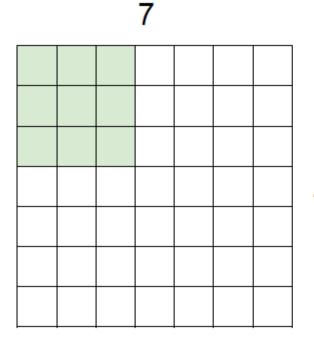


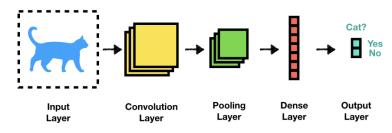




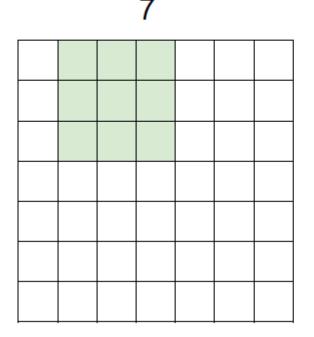


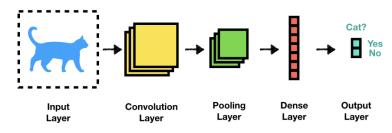
A closer look at spatial dimensions:



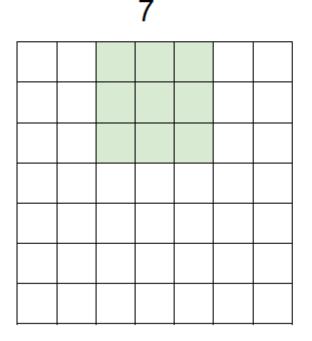


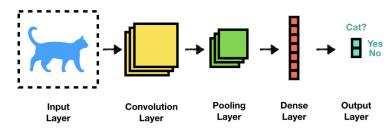
A closer look at spatial dimensions:



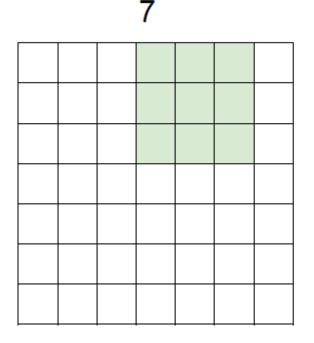


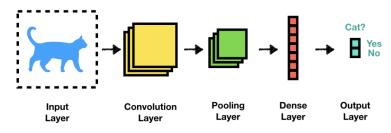
A closer look at spatial dimensions:



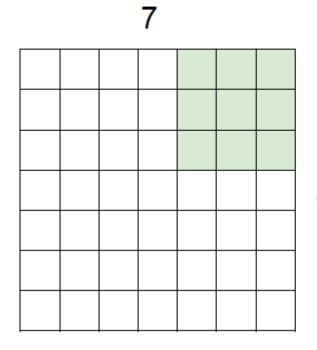


A closer look at spatial dimensions:



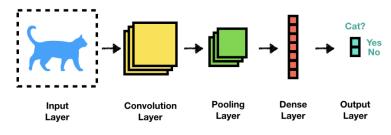


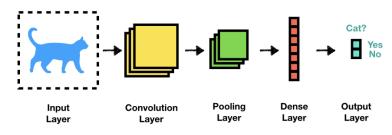
A closer look at spatial dimensions:



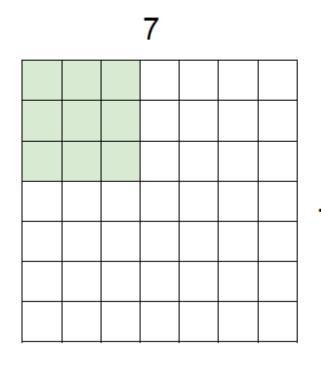
7x7 input (spatially) assume 3x3 filter

=> 5x5 output

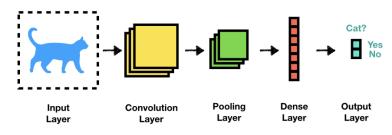




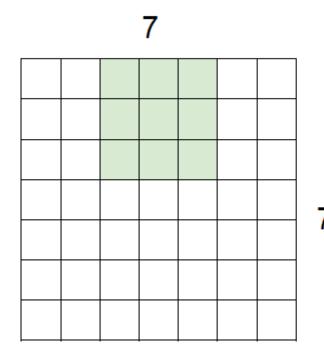
A closer look at spatial dimensions:



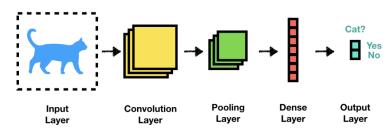
7x7 input (spatially) assume 3x3 filter applied with stride 2



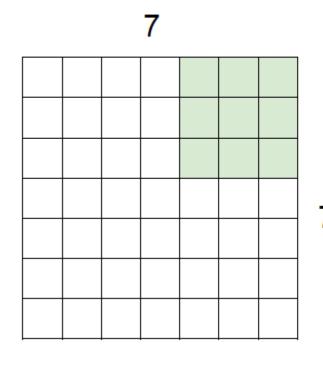
A closer look at spatial dimensions:



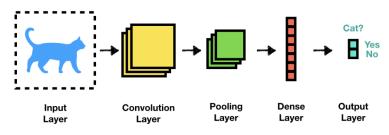
7x7 input (spatially) assume 3x3 filter applied with stride 2



A closer look at spatial dimensions:



7x7 input (spatially)
assume 3x3 filter
applied with stride 2
=> 3x3 output!

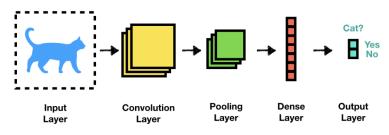


In practice: Common to zero pad the border

0	0	0	0	0	0		
0							
0							
0							
0							

e.g. input 7x7
3x3 filter, applied with stride 1
pad with 1 pixel border => what is the output?

```
(recall:)
(N - F) / stride + 1
```



In practice: Common to zero pad the border

0	0	0	0	0	0		
0							
0							
0							
0							

e.g. input 7x7

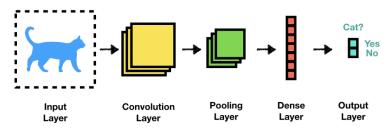
3x3 filter, applied with stride 1

pad with 1 pixel border => what is the output?

7x7 output!

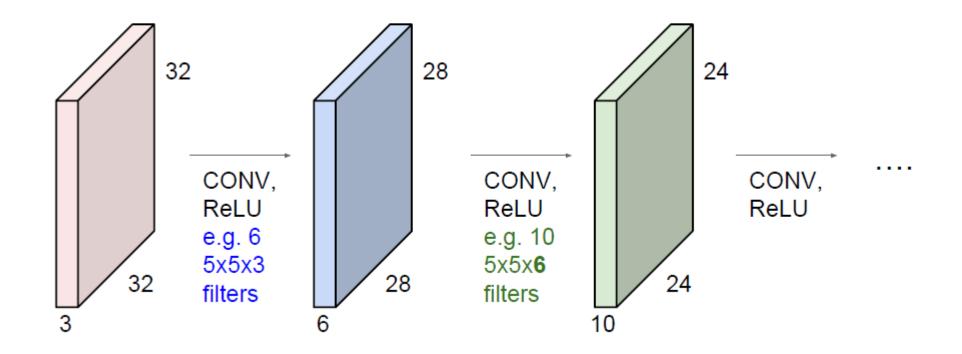
in general, common to see CONV layers with stride 1, filters of size FxF, and zero-padding with (F-1)/2. (will preserve size spatially)

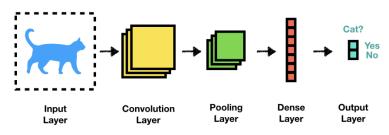
e.g. F = 3 => zero pad with 1 F = 5 => zero pad with 2 F = 7 => zero pad with 3



Remember back to...

E.g. 32x32 input convolved repeatedly with 5x5 filters shrinks volumes spatially! (32 -> 28 -> 24 ...). Shrinking too fast is not good, doesn't work well.



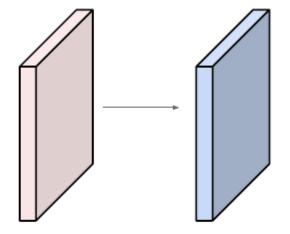


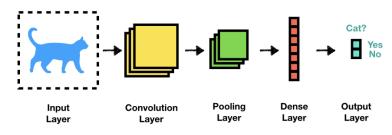
Examples time:

Input volume: 32x32x3

10 5x5 filters with stride 1, pad 2

Output volume size: ?

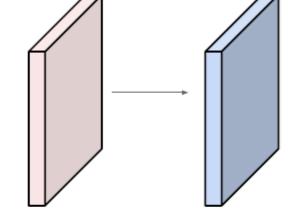




Examples time:

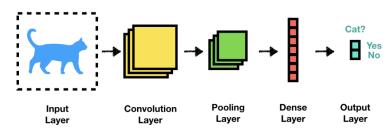
Input volume: 32x32x3

10 5x5 filters with stride 1, pad 2



Output volume size:

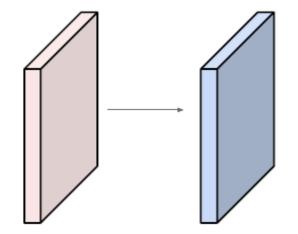
(32+2*2-5)/1+1 = 32 spatially, so 32x32x10



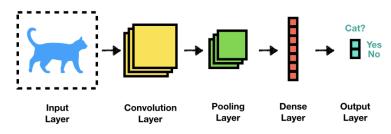
Examples time:

Input volume: 32x32x3

10 5x5 filters with stride 1, pad 2



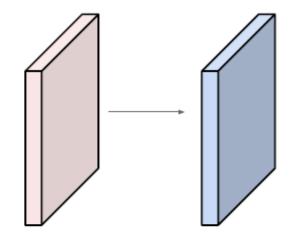
Number of parameters in this layer?



Examples time:

Input volume: 32x32x3

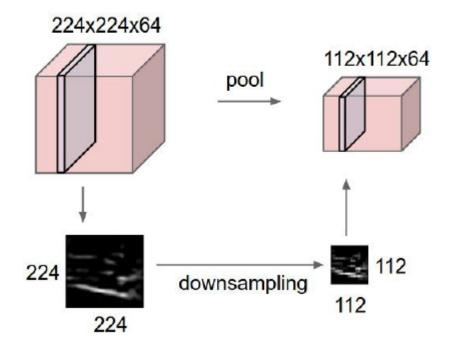
10 5x5 filters with stride 1, pad 2

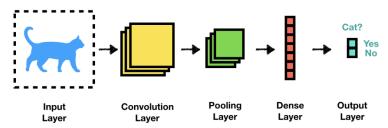


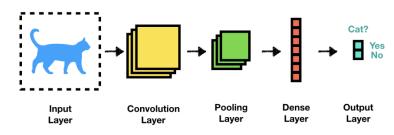
Number of parameters in this layer? each filter has 5*5*3 + 1 = 76 params (+1 for bias) => 76*10 = 760

Pooling layer

- makes the representations smaller and more manageable
- operates over each activation map independently:







MAX POOLING

Single depth slice

X		1	1	2	4
		5	6	7	8
		3	2	1	0
		1	2	3	4
	_				

max pool with 2x2 filters and stride 2

6	8	
3	4	