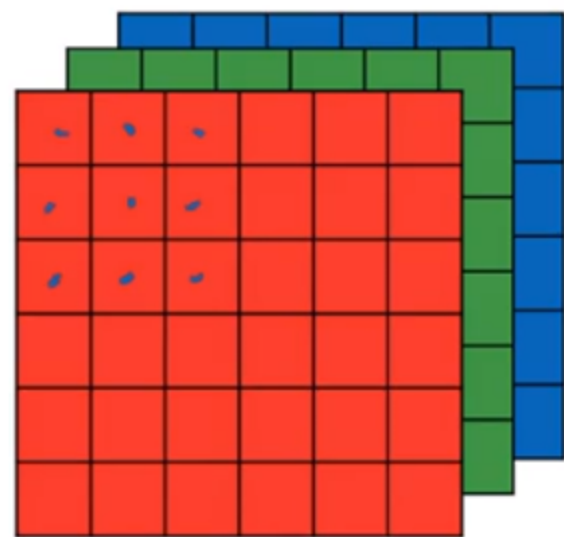


# Example of a layer



$6 \times 6 \times 3$

$a^{[1]}$

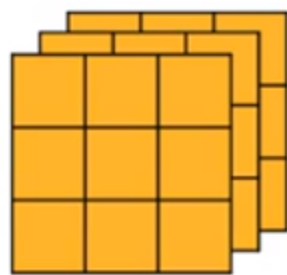
$$z^{[1]} = W^{[1]} a^{[0]} + b^{[1]}$$

$$a^{[1]} = g(z^{[1]})$$

$*$

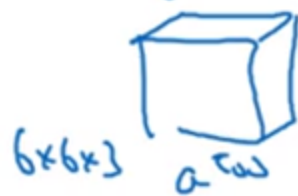


$3 \times 3 \times 3$



$3 \times 3 \times 3$

$W^{[1]}$

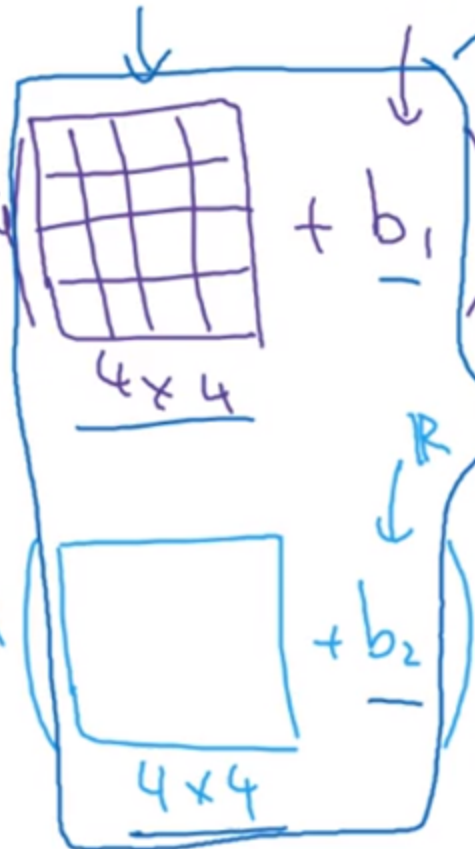


$6 \times 6 \times 3$

$a^{[0]}$

$\rightarrow \text{ReLU}$

$\rightarrow \text{ReLU}$



$4 \times 4$

$4 \times 4$



$4 \times 4$



$4 \times 4$



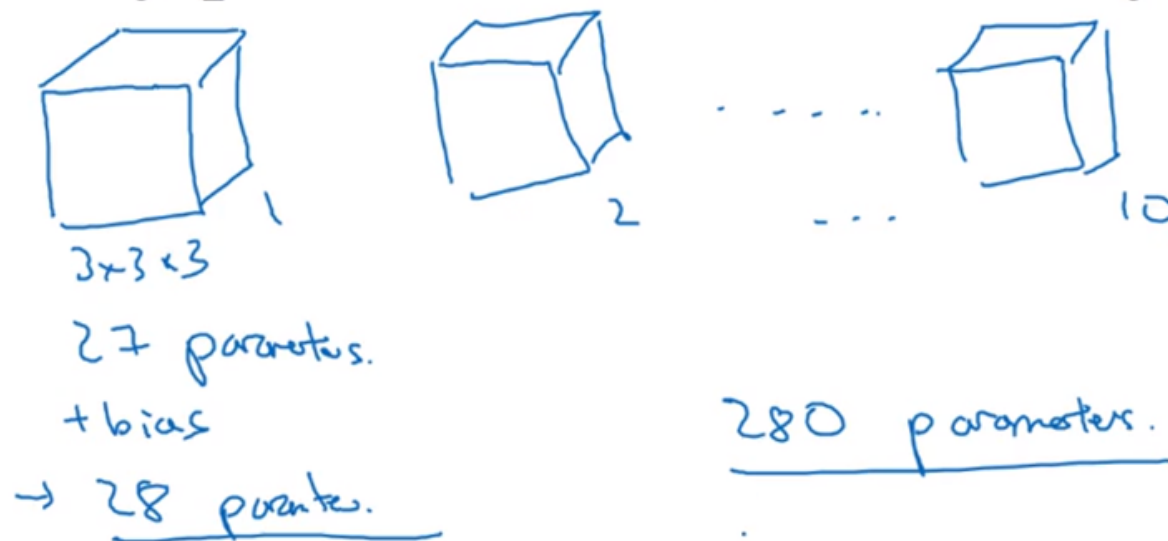
$4 \times 4 \times 2$

$a^{[1]}$

$4 \times 4 \times 10$

# Number of parameters in one layer

If you have 10 filters that are  $3 \times 3 \times 3$  in one layer of a neural network, how many parameters does that layer have?



Andrew Ng

So this is really one property of convolutional neural network that makes them less prone to over fitting, that you could, so last you've learned 10 feature detectors that work you could apply this to even large images and the number of parameters still remain fixed and relatively small, as 280 in this example

# Summary of notation

If layer l is a convolution layer:

$f^{[l]}$  = filter size

$p^{[l]}$  = padding

$s^{[l]}$  = stride

$n_c^{[l]}$  = number of filters

→ Each filter is:  $f^{[l]} \times f^{[l]} \times n_c^{[l-1]}$

Activations:  $a^{[l]} \rightarrow n_H^{[l]} \times n_W^{[l]} \times n_c^{[l]}$

Weights:  $f^{[l]} \times f^{[l]} \times n_c^{[l-1]} \times n_c^{[l]}$

bias:  $n_c^{[l]} = (1, 1, 1, n_c^{[l]})$  ← #filters in layer l.

Input:  $n_H^{[l-1]} \times n_W^{[l-1]} \times n_c^{[l-1]}$  ←  
Output:  $n_H^{[l]} \times n_W^{[l]} \times n_c^{[l]}$  ←

$$n_H^{[l]} = \left\lfloor \frac{n_H^{[l-1]} + 2p^{[l]} - f^{[l]}}{s^{[l]}} + 1 \right\rfloor$$

$$A^{[l]} \rightarrow m \times \underbrace{n_H^{[l]} \times n_W^{[l]} \times n_c^{[l]}}_{n_c^{[l]} \times n_H^{[l]} \times n_W^{[l]}}$$