

Machine Learning Using Tensorflow

Week 6: Convolution Neural Network

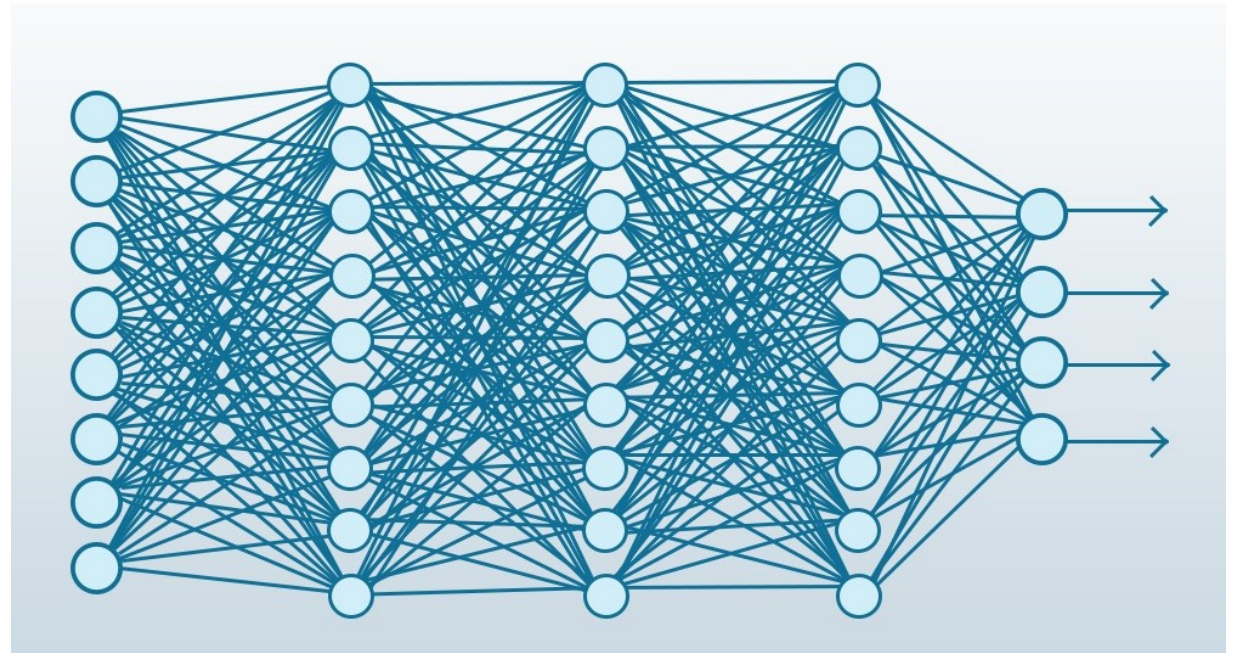
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UC Davis



How many parameters?



256 x 256 = 65,536 pixels



input
65,536

hidden1
x 32738

hidden2
x 16384

hidden3
x 8192

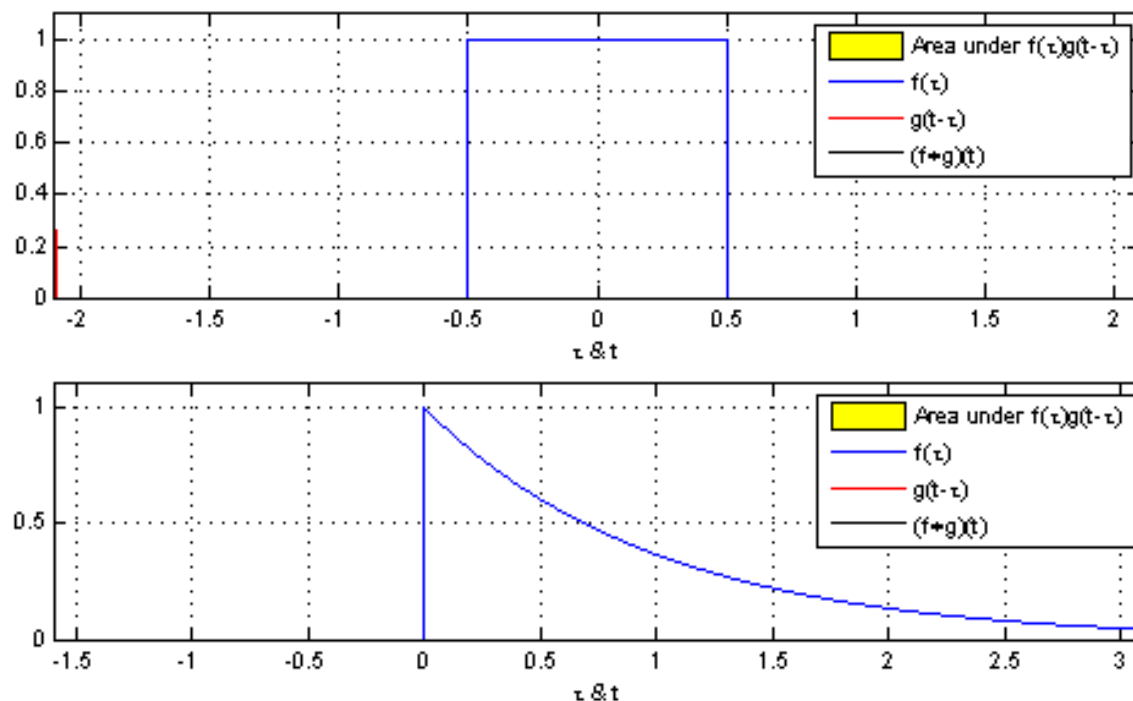
Output
x 10

Trainable Parameters ~ 3×10^{18} ! Ridiculous !

What is Convolution?

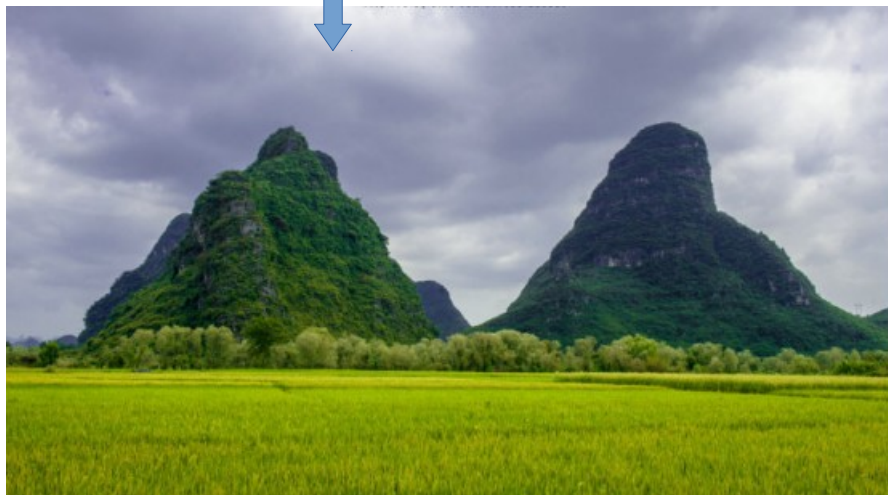
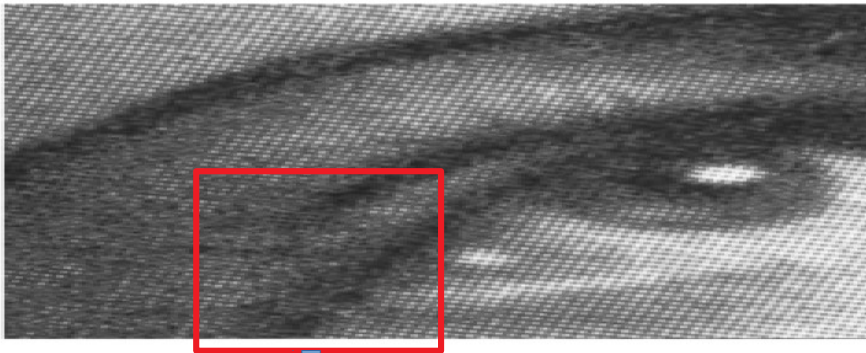
$$(f * g)(n) = \int_{-\infty}^{\infty} f(\tau)g(n - \tau)d\tau$$

$$(f * g)(n) = \sum_{\tau=-\infty}^{\infty} f(\tau)g(n - \tau)$$



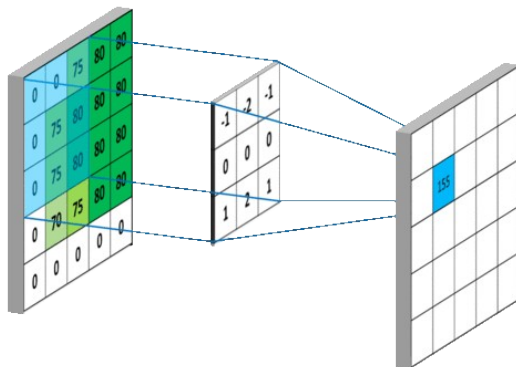
Weighted sum of a function !

Use it to smooth an image



$$\begin{aligned} f &= \begin{bmatrix} a_{0,0} & a_{0,1} & a_{0,2} \\ a_{1,0} & a_{1,1} & a_{1,2} \\ a_{2,0} & a_{2,1} & a_{2,2} \end{bmatrix} \\ \begin{bmatrix} a_{0,0} & a_{0,1} & a_{0,2} & \cdots & a_{0,n} \\ a_{1,0} & a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ a_{2,0} & a_{2,1} & a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m,0} & a_{m,1} & a_{m,2} & \cdots & a_{m,n} \end{bmatrix} &\Rightarrow \begin{bmatrix} c_{0,0} & c_{0,1} & c_{0,2} & \cdots & c_{1,n} \\ c_{1,0} & c_{1,1} & c_{1,2} & \cdots & c_{1,n} \\ c_{2,0} & c_{2,1} & c_{2,2} & \cdots & c_{2,n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ c_{m,0} & c_{m,1} & c_{m,2} & \cdots & c_{m,n} \end{bmatrix} \\ &\quad \uparrow \\ &\quad c_{1,1} = f * g \\ &\quad \uparrow \\ g &= \begin{bmatrix} \frac{1}{9} & \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} & \frac{1}{9} \\ \frac{1}{9} & \frac{1}{9} & \frac{1}{9} \end{bmatrix} \end{aligned}$$

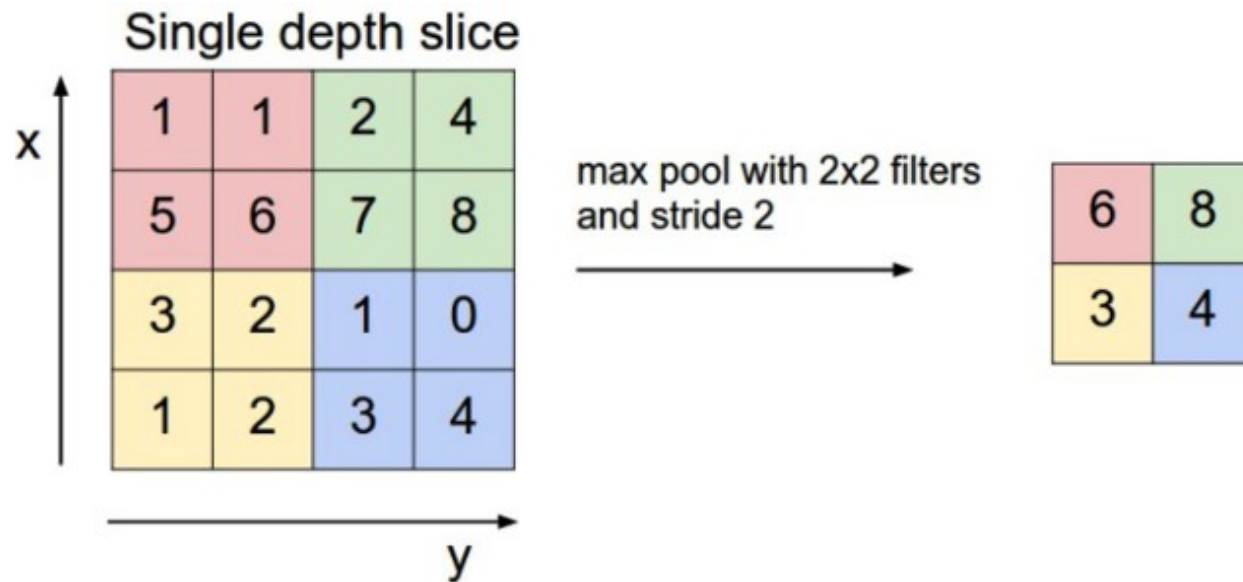
Continuous Convolution



Input Volume (+pad 1) (7x7x3)	Filter W0 (3x3x3)	Filter W1 (3x3x3)	Output Volume (3x3x2)
$x[:, :, 0]$	$w0[:, :, 0]$	$w1[:, :, 0]$	$o[:, :, 0]$
0 0 0 0 0 0 0	-1 1 0	1 1 -1	6 7 5
0 0 1 1 0 2 0	0 1 0	-1 -1 1	3 -1 -1
0 2 2 2 2 1 0	0 1 1	0 -1 1	2 -1 4
0 1 0 0 2 0 0	-1 -1 0	0 1 0	2 -5 -8
0 0 1 1 0 0 0	0 0 0	-1 0 -1	1 -4 -4
0 1 2 0 0 2 0	0 -1 0	-1 1 0	0 -5 -5
0 0 0 0 0 0 0			
$x[:, :, 1]$	$w0[:, :, 1]$	$w1[:, :, 1]$	$o[:, :, 1]$
0 0 0 0 0 0 0	0 0 1	-1 0 0	
0 1 0 2 2 0 0	0 1 0	-1 0 1	
0 0 0 0 2 0 0	1 -1 -1	-1 0 0	
0 1 2 1 2 1 0			
0 1 0 0 0 0 0			
0 1 2 1 1 1 0			
0 0 0 0 0 0 0			
$x[:, :, 2]$	Bias $b0$ (1x1x1)	Bias $b1$ (1x1x1)	
0 0 0 0 0 0 0	$b0[:, :, 0]$	$b1[:, :, 0]$	
0 2 1 2 0 0 0	1	0	
0 1 0 0 1 0 0			
0 0 2 1 0 1 0			
0 0 1 2 2 2 0			
0 2 1 0 0 1 0			
0 0 0 0 0 0 0			

toggle movement

Max Pooling

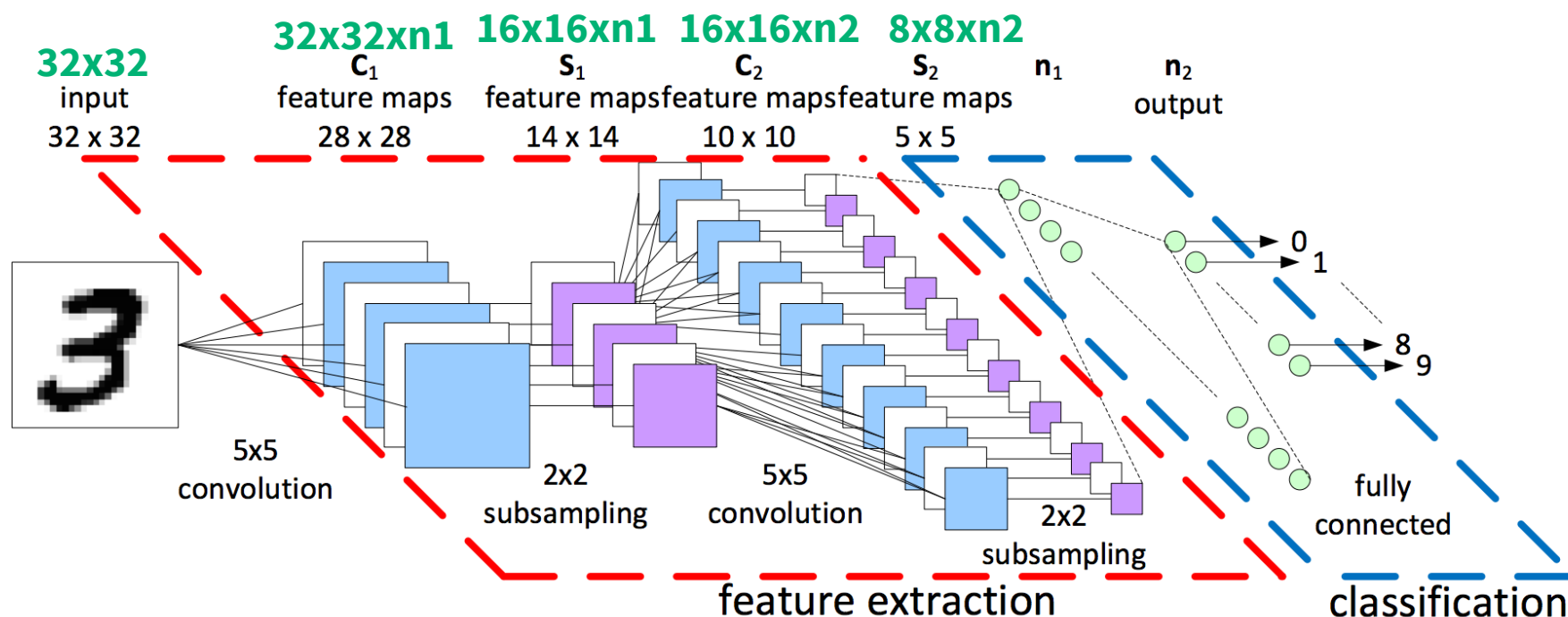


Use the maximal value to represent the main feature of a block

Procedure of CNN

Check this video released by Google:

- <https://www.youtube.com/watch?v=jajksuQW4mc>



All image block shares the same “filter lens”, parameters are highly reduced!

Image size: 32x32 → 8x8xn2 (width → depth)

Why it works?

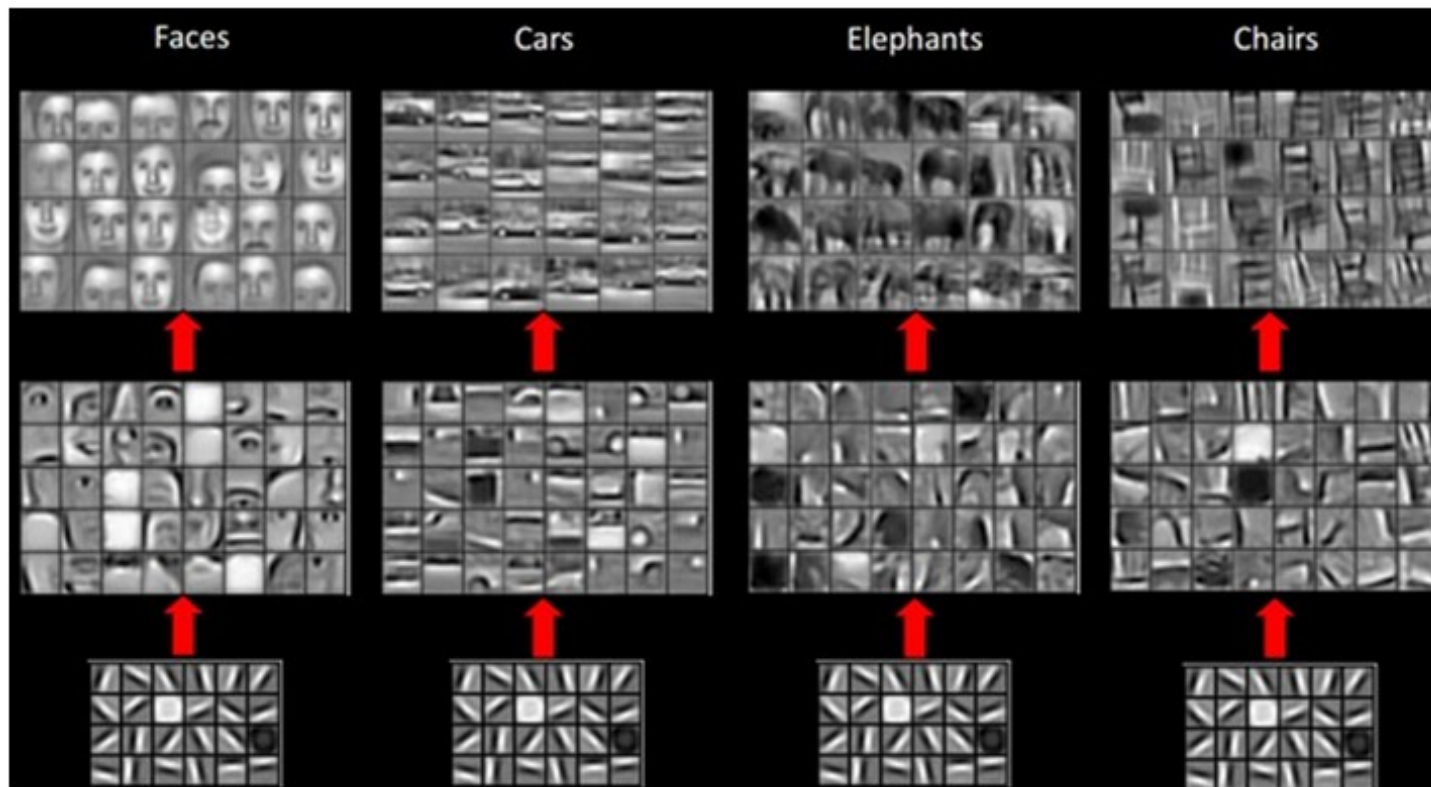
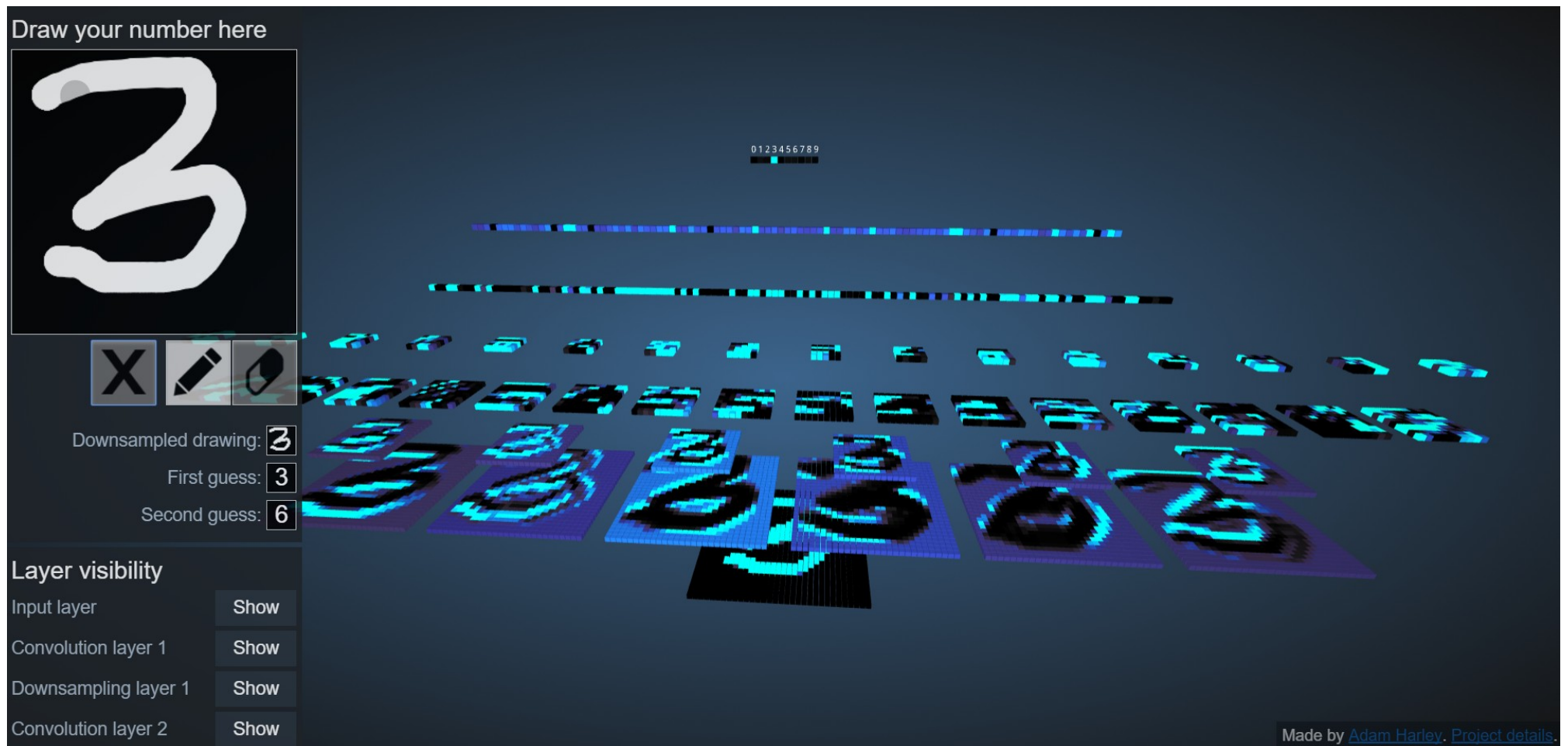


Image = local features merge to global feature !
(actually not only image but almost all kinds of information)

Convolution as a “filter”

<http://scs.ryerson.ca/~aharley/vis/conv/>



Exercise

code06_ex1_v1:

Using TensorFlow to perform CNN

code06_ex1_v2:

Using TensorFlow to perform CNN

Using t-SNE to cluster the output of flatten layer

code06_ex2_v2:

Using Keras to perform CNN