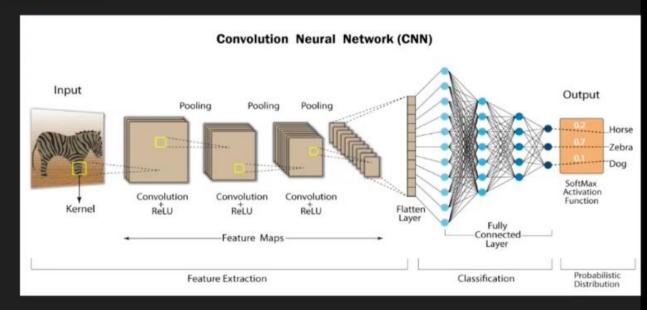
What is a CNN?

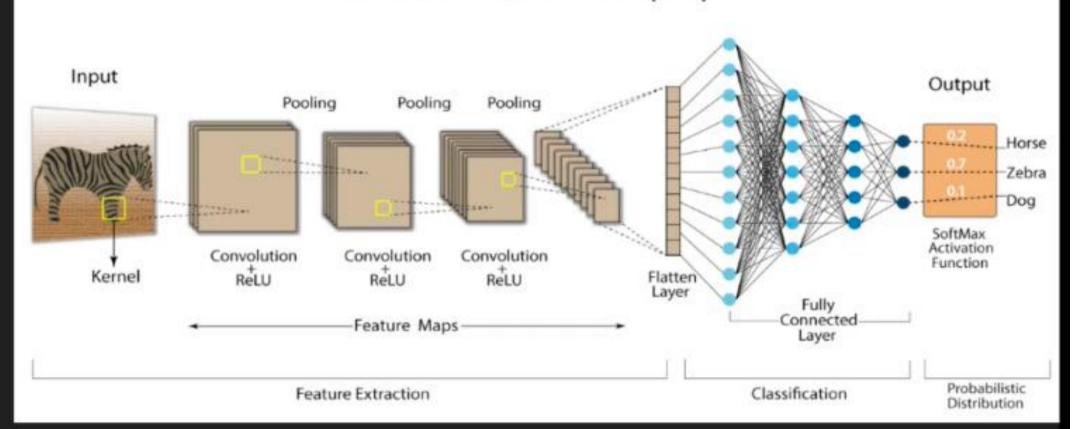
17 August 2022

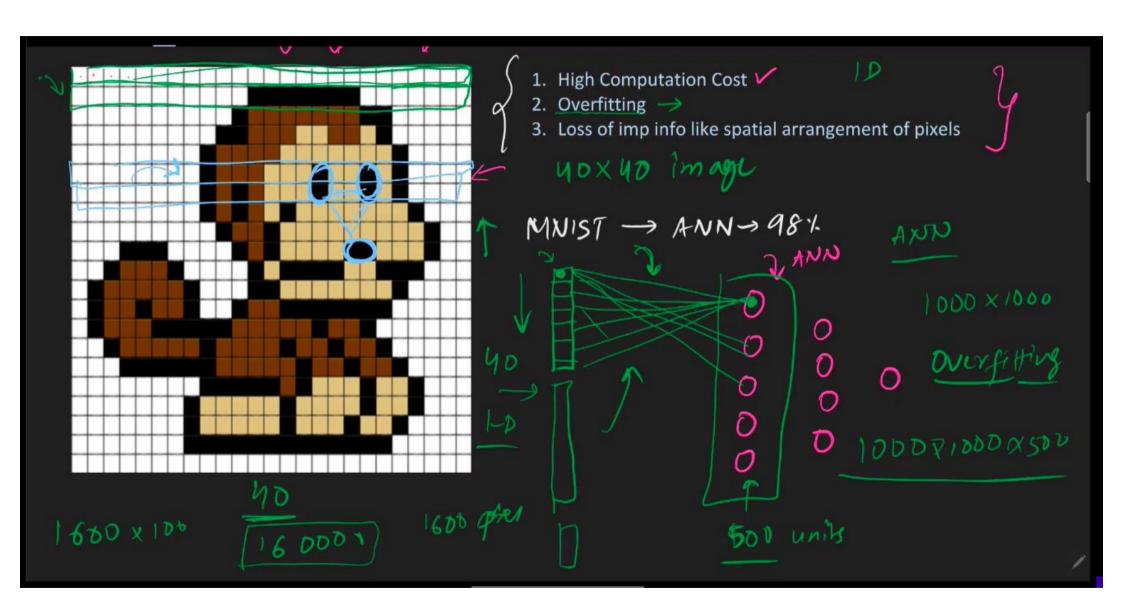
06:47

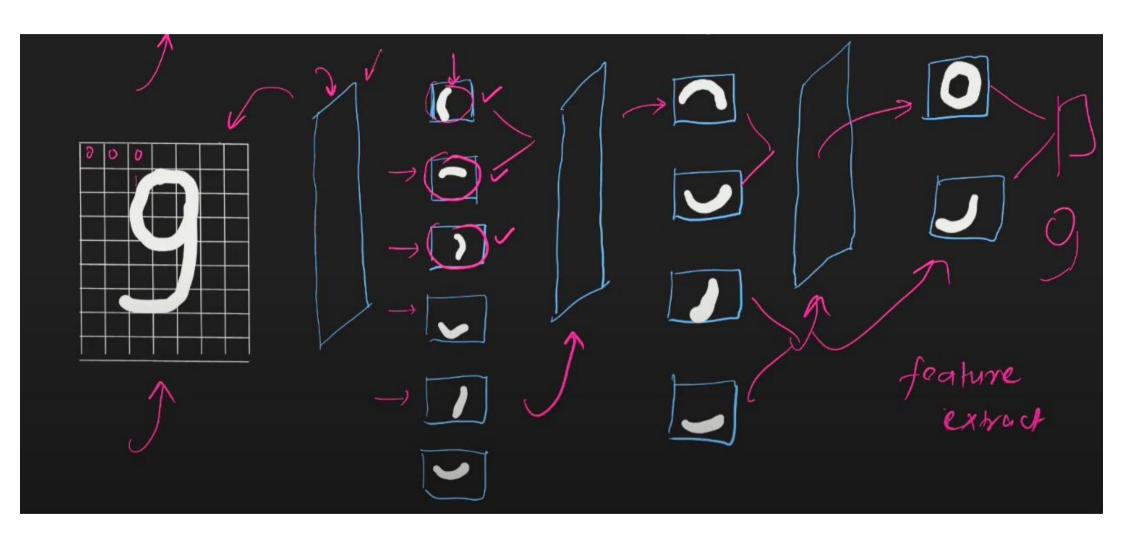
Convolutional neural networks, also known as convnet, or CNNs, are a special kind of neural network for processing data that has a known grid-like topology like time series data(1D) or images(2D).



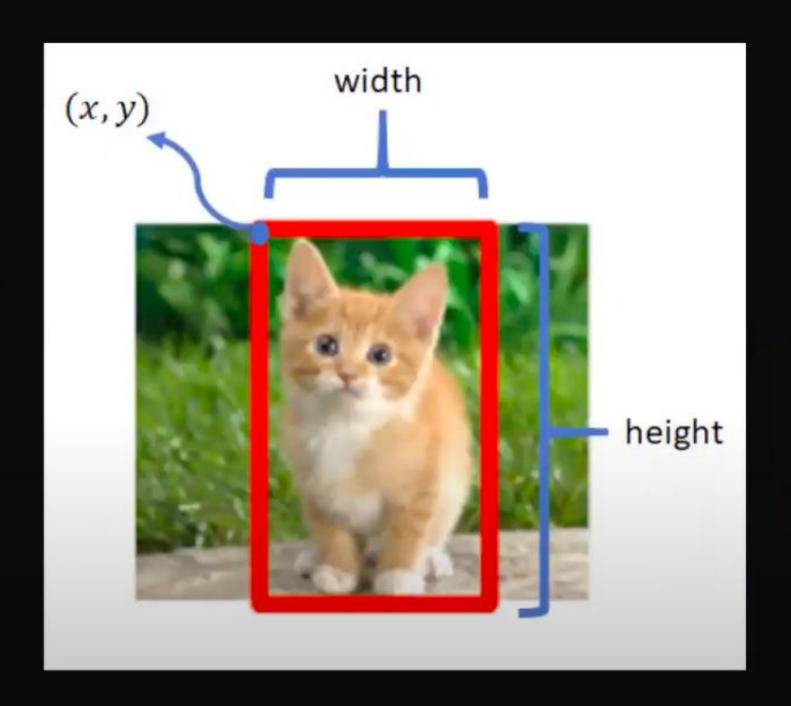
Convolution Neural Network (CNN)

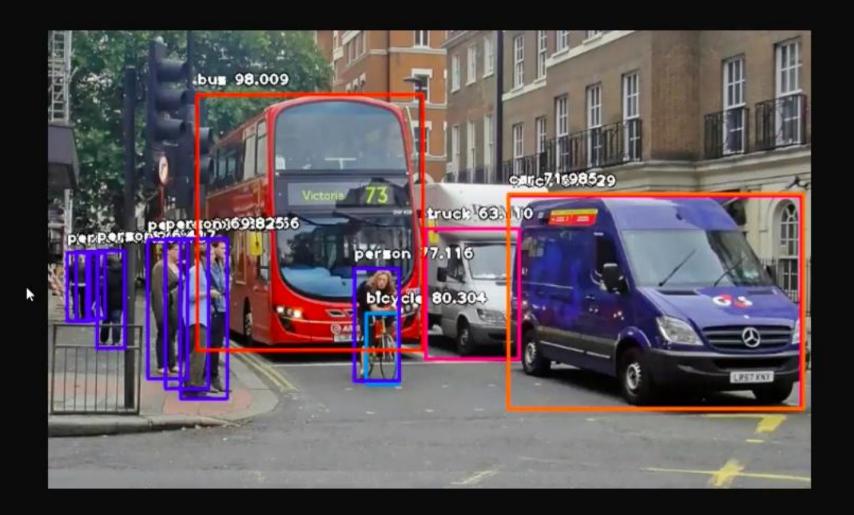


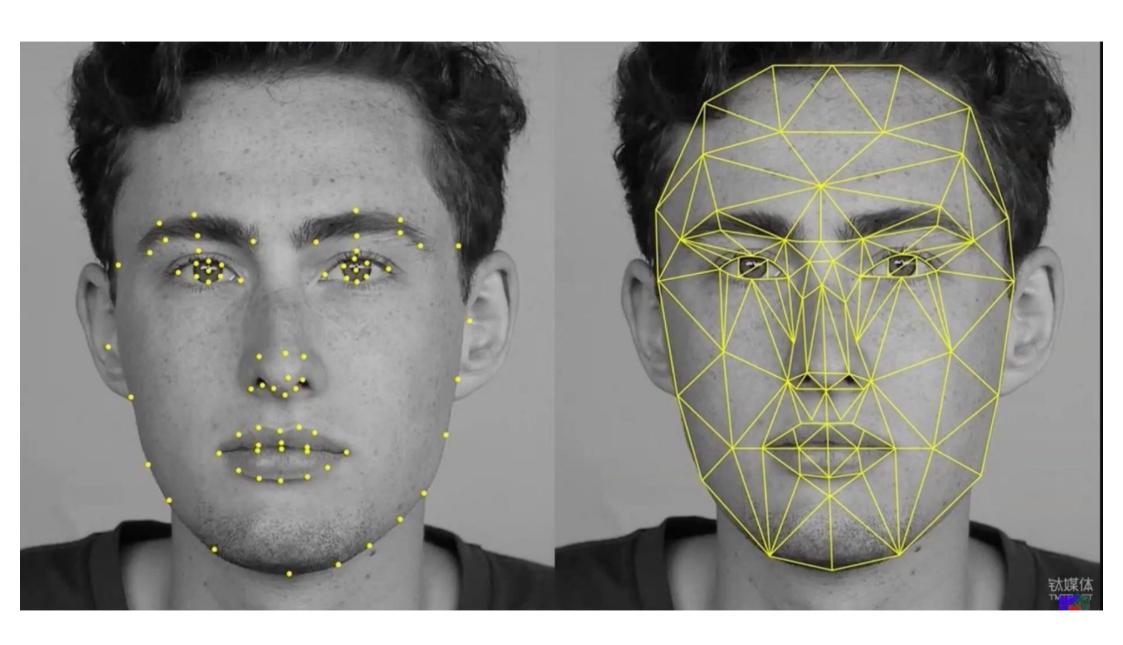


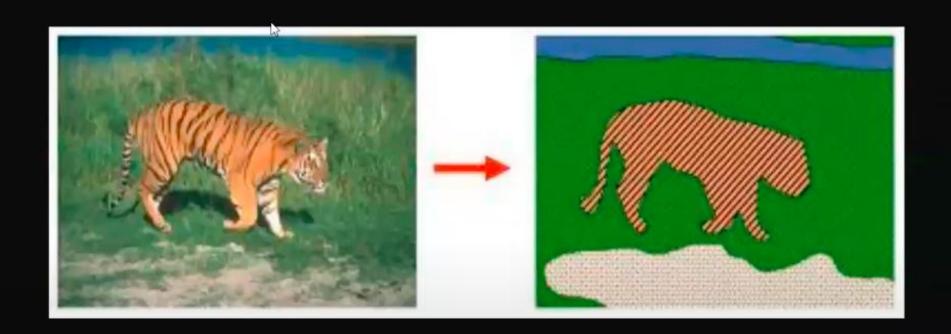












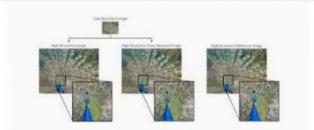












Increase Image Resolution Using Deep ... mathworks.com



Deep Learning based image Super ... cv-tricks.com

semanticscholar.org

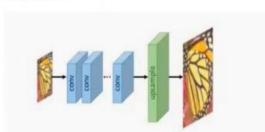
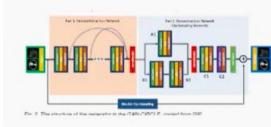
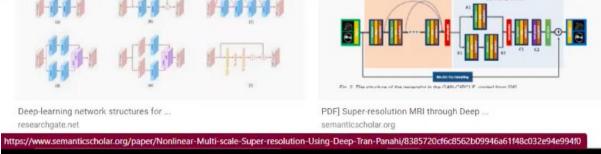


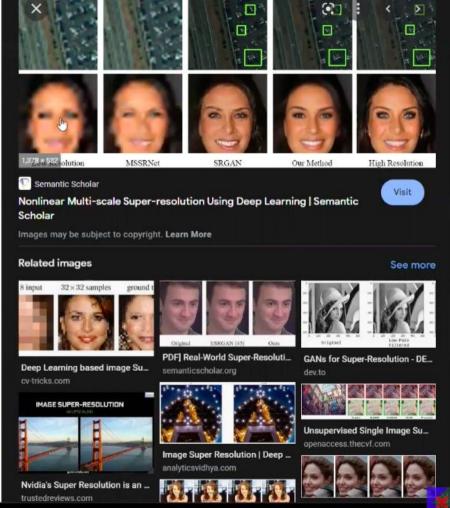
Image Super Resolution | Deep Learning ... analyticsvidhya.com

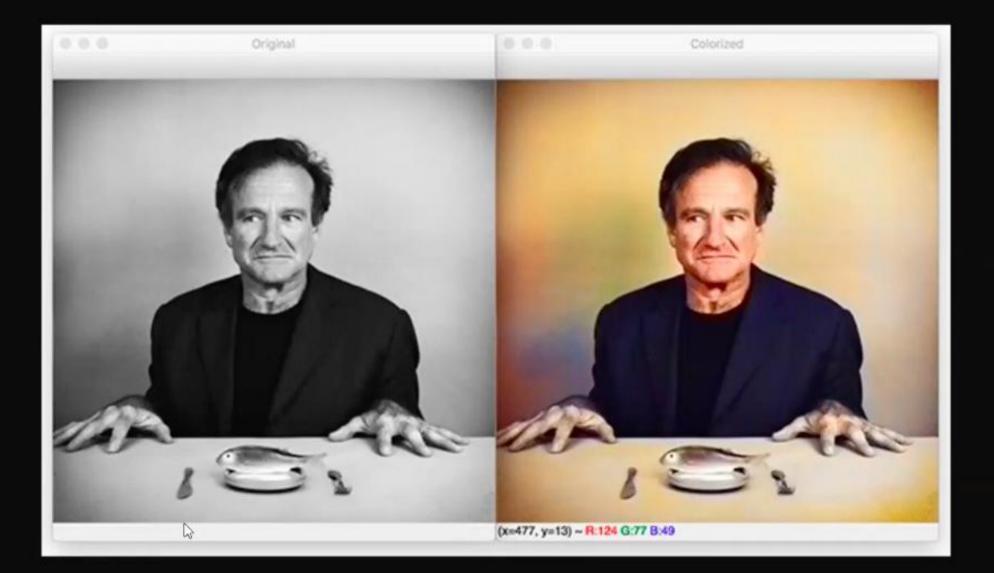




Deep-learning network structures for ... researchgate.net







5:37 [] 🕱 🔻 🕏 53% 🖥

D36 CNN vs Visual Cortex HOME INSERT DRAW

VIEW









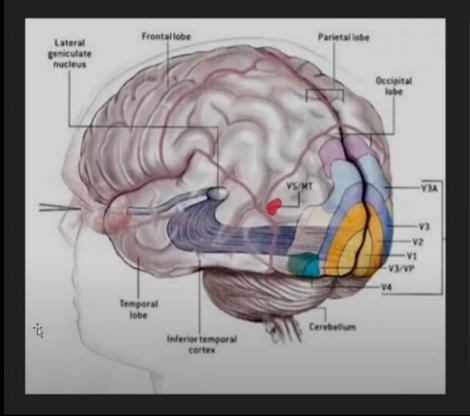


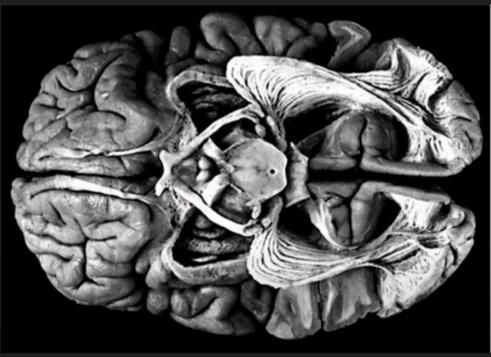




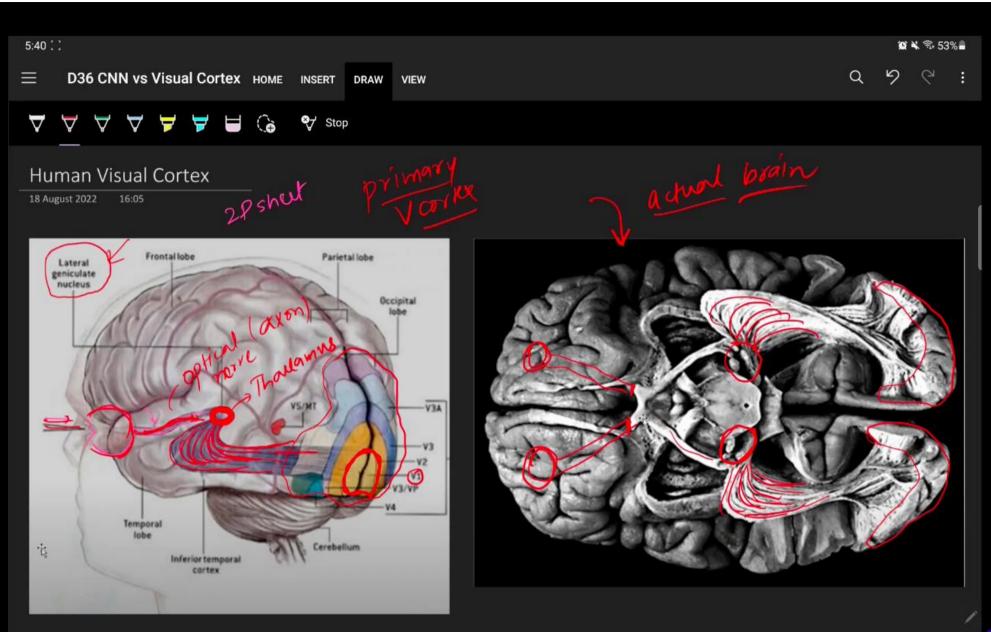
Human Visual Cortex

18 August 2022 16:05

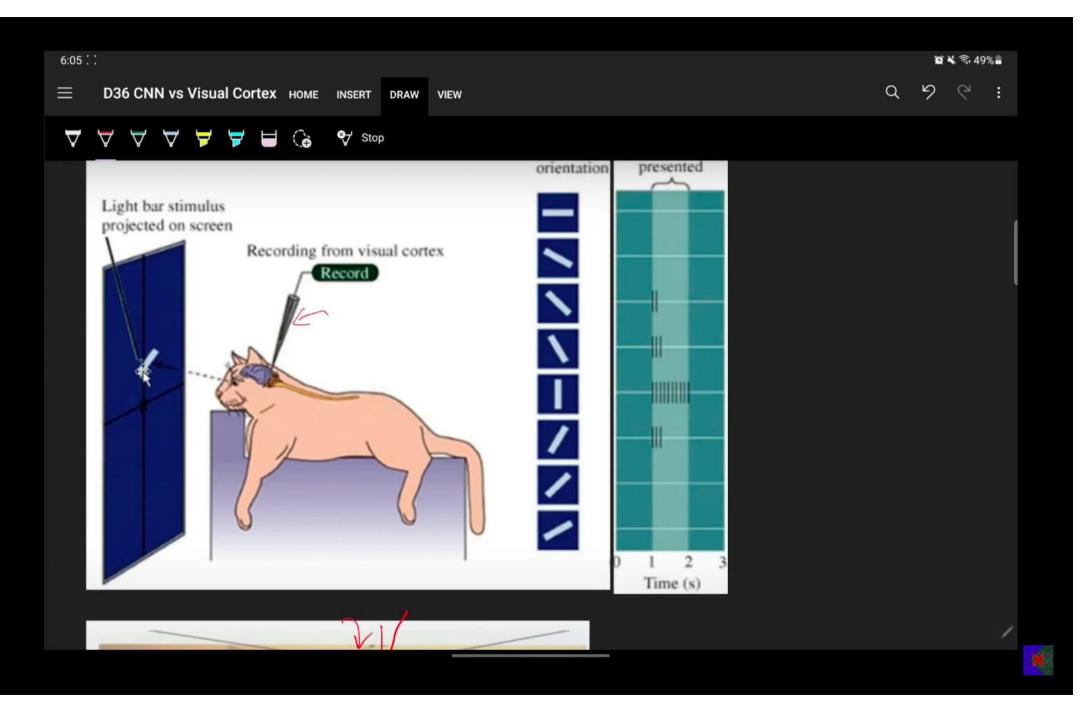


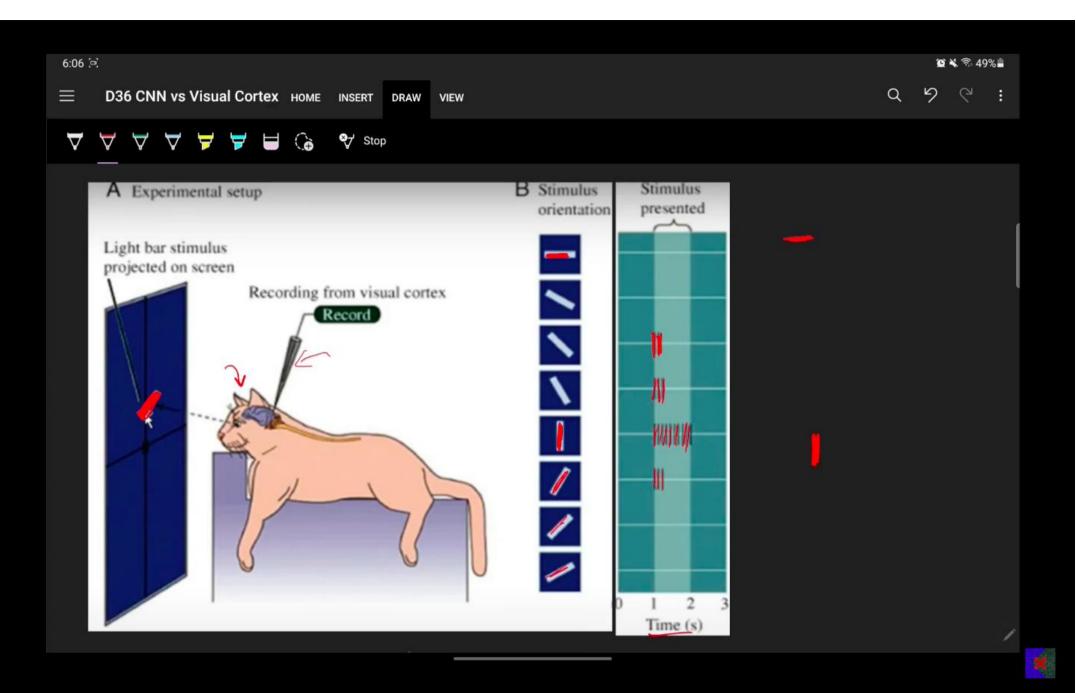


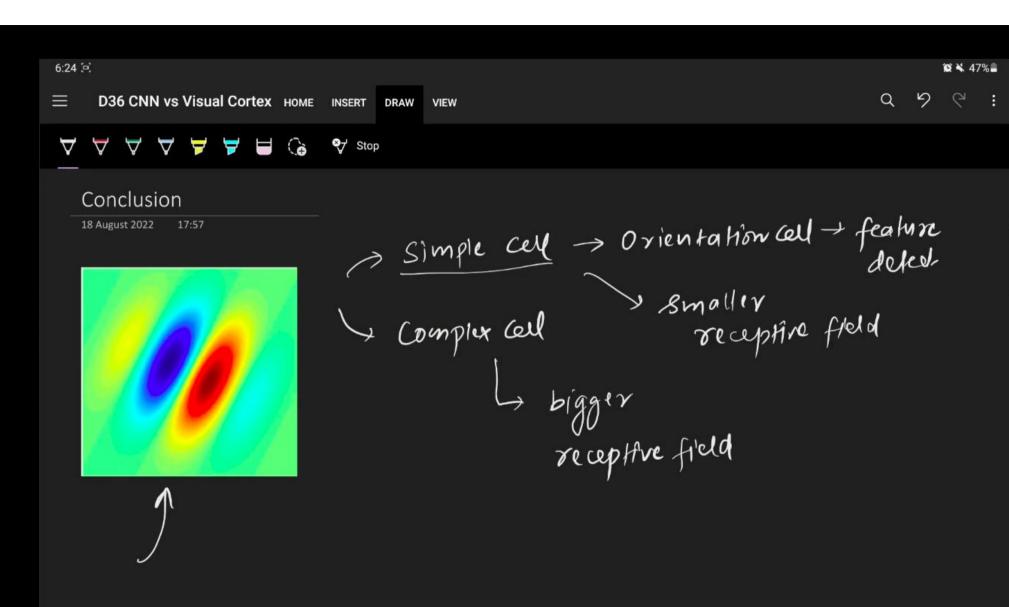








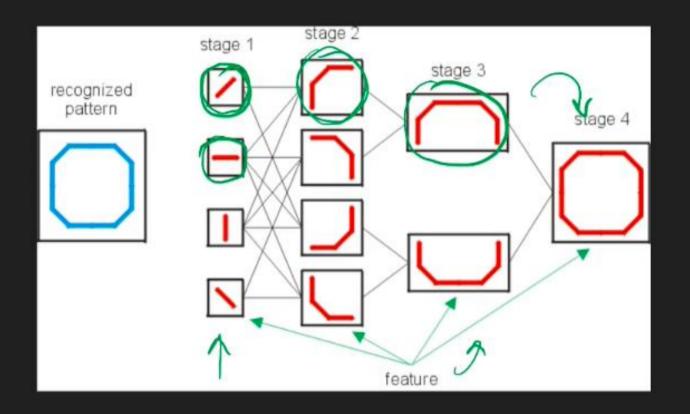




Development

18 August 2022 1

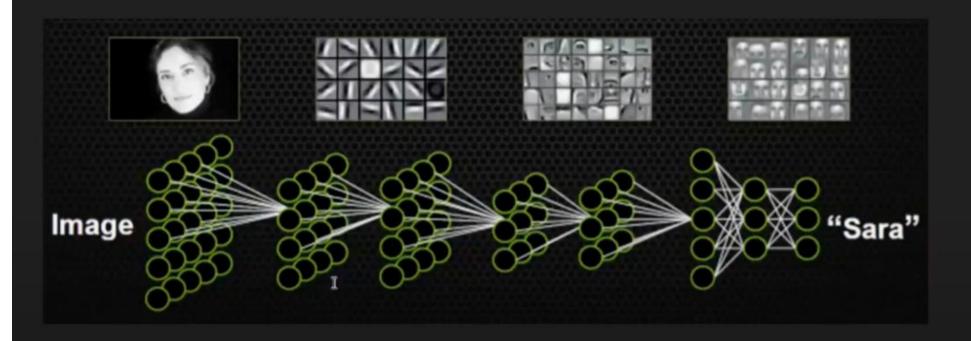
16:15



Introduction

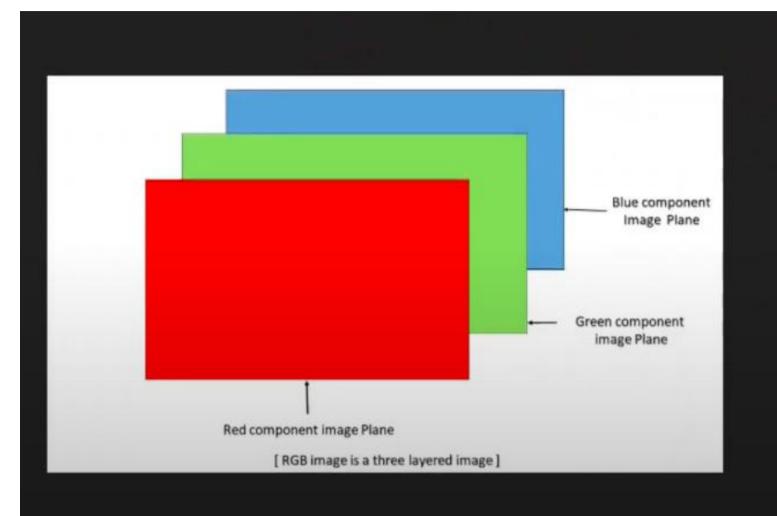
19 August 2022

16:45





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52	82	53	E2	0.0	80		2.5	1.8	12	83	0,0	8	60	88	88	82	60	魒	05	1.0	10	0.8	ac	82	0.3	80	88
00)	0.0	60	50	en.	no	na.	0.0	1.0	0.0	ħΩ	as	0.0	no	0,0		0.0	0.0	NA.	63	1.0	1.0	08	αn	En.	68	na	(qm)
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0.0	50	00	篋	0.0	nn	03	3.0	10	0.8	0.7	01	(0.3)	0.7	ON	1.0	0.0	0.7	82	0.0	0.0	8	00	me	88	0.0	na	88
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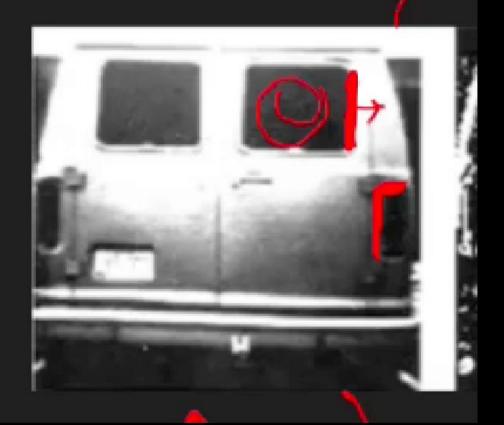
Edge Detection (Convolution Operation)

19 August 2022

16:53

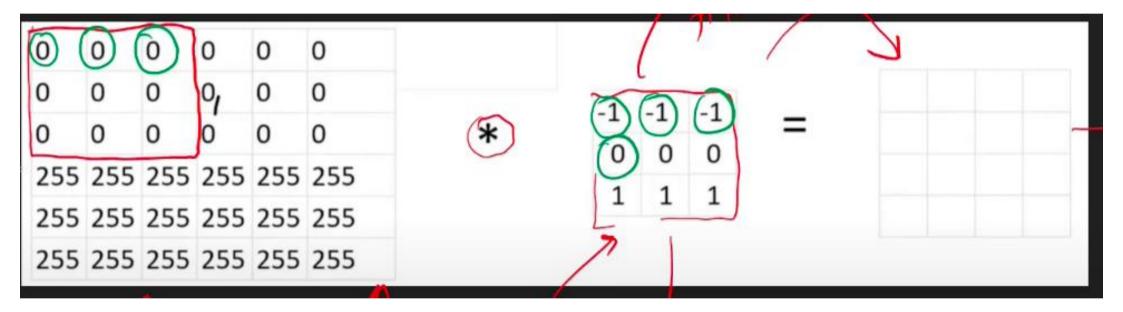


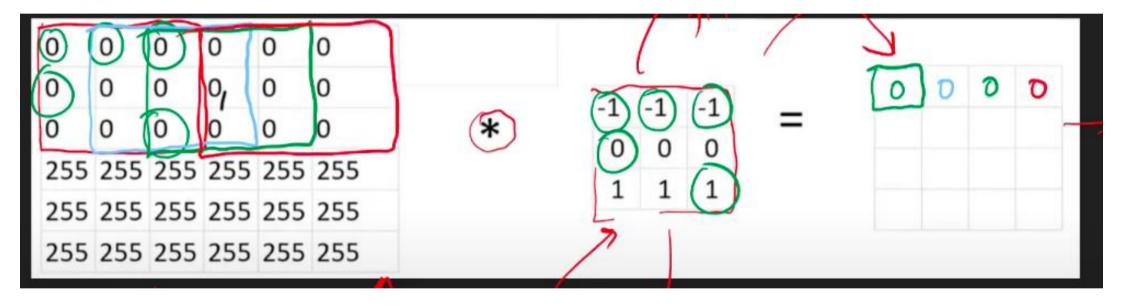


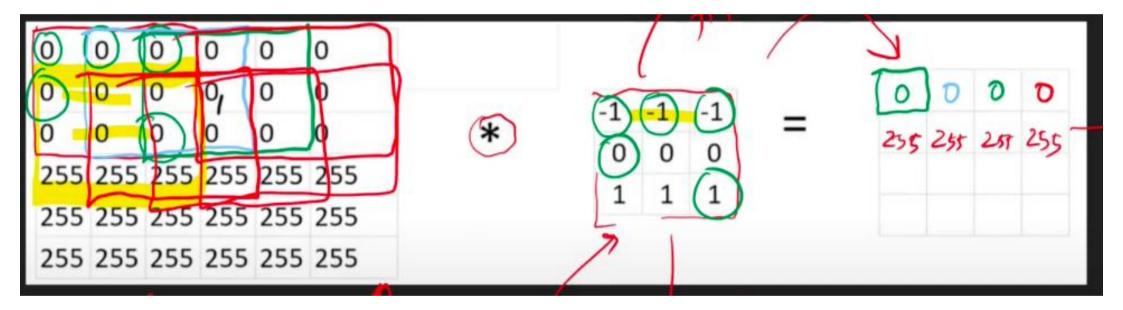


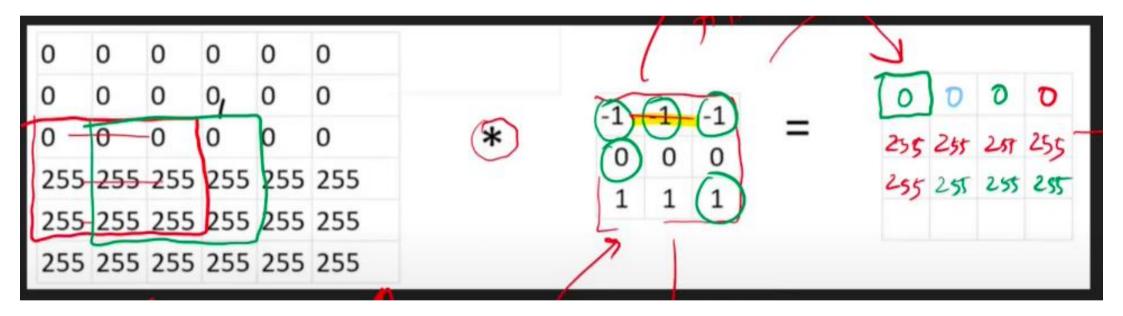
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0	0	0	0	0	0						
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à.											
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255	255	255	255	255	255		1	1	1		
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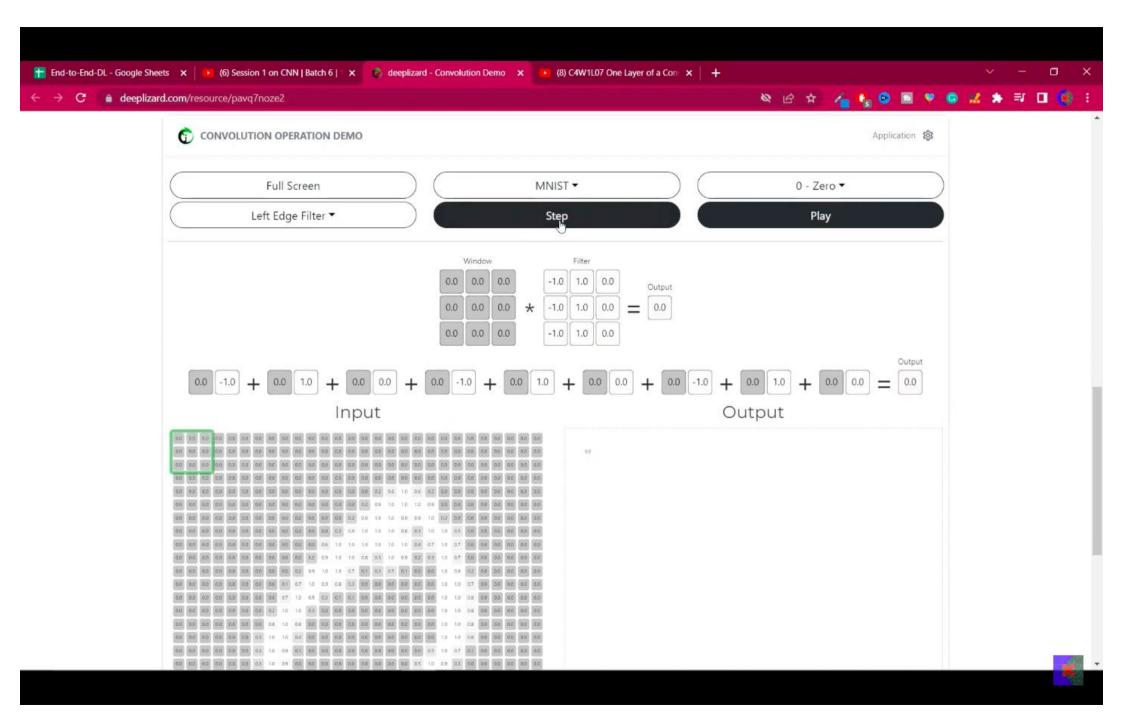


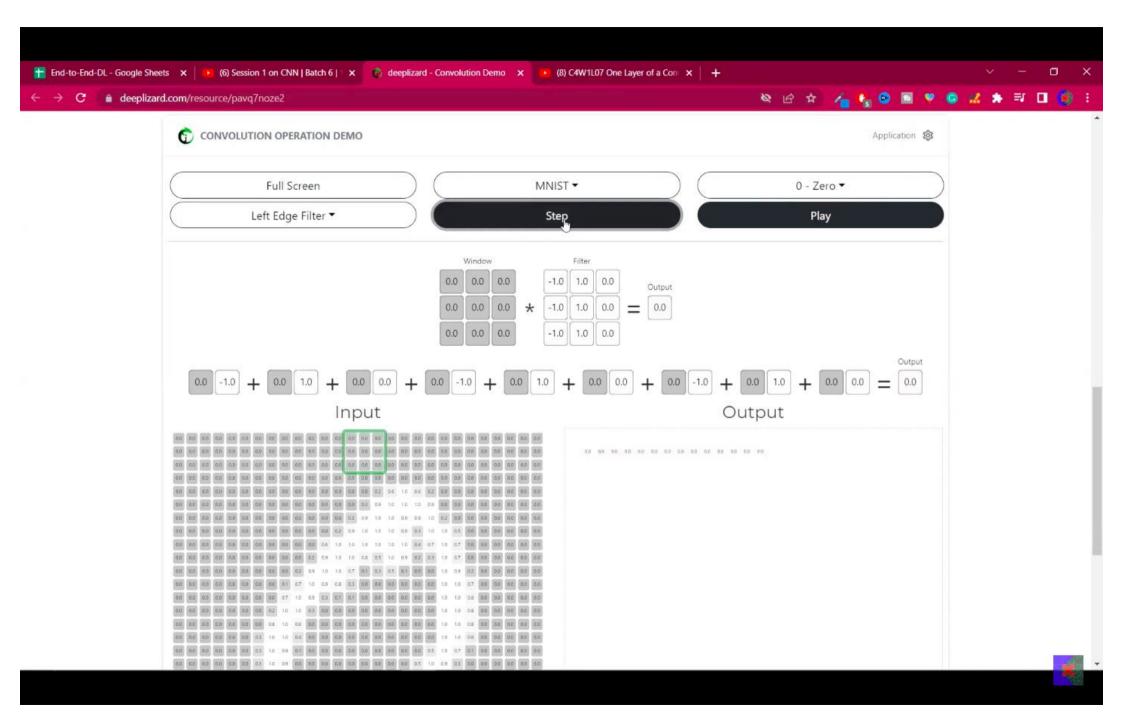


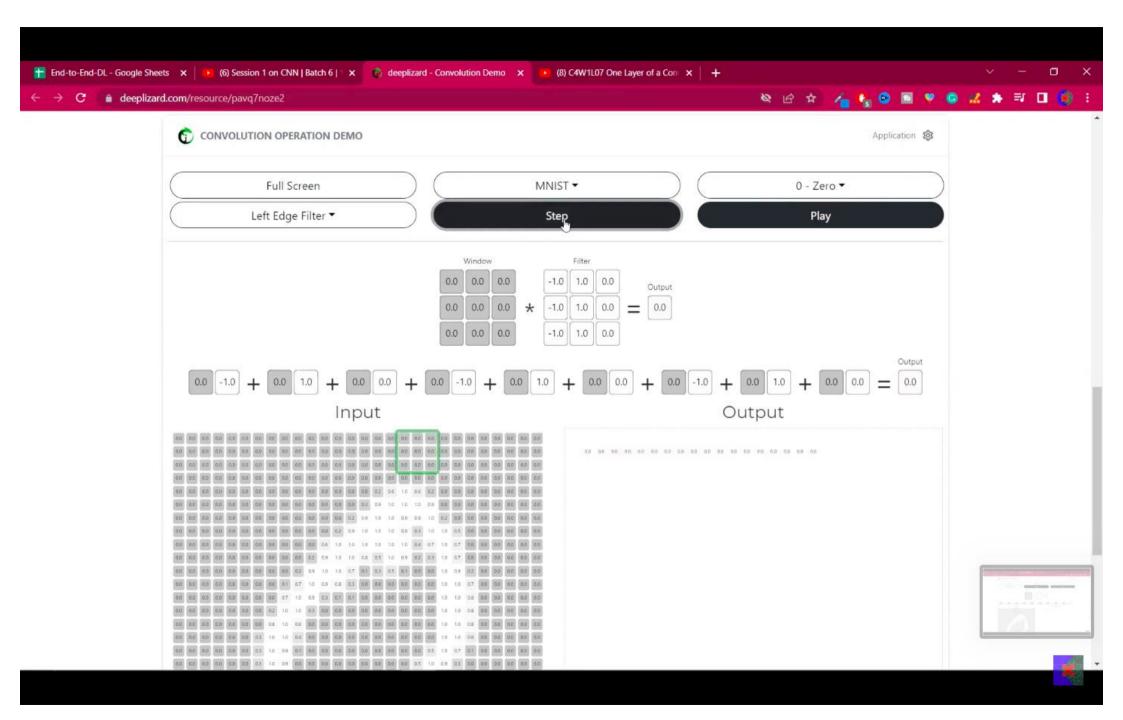


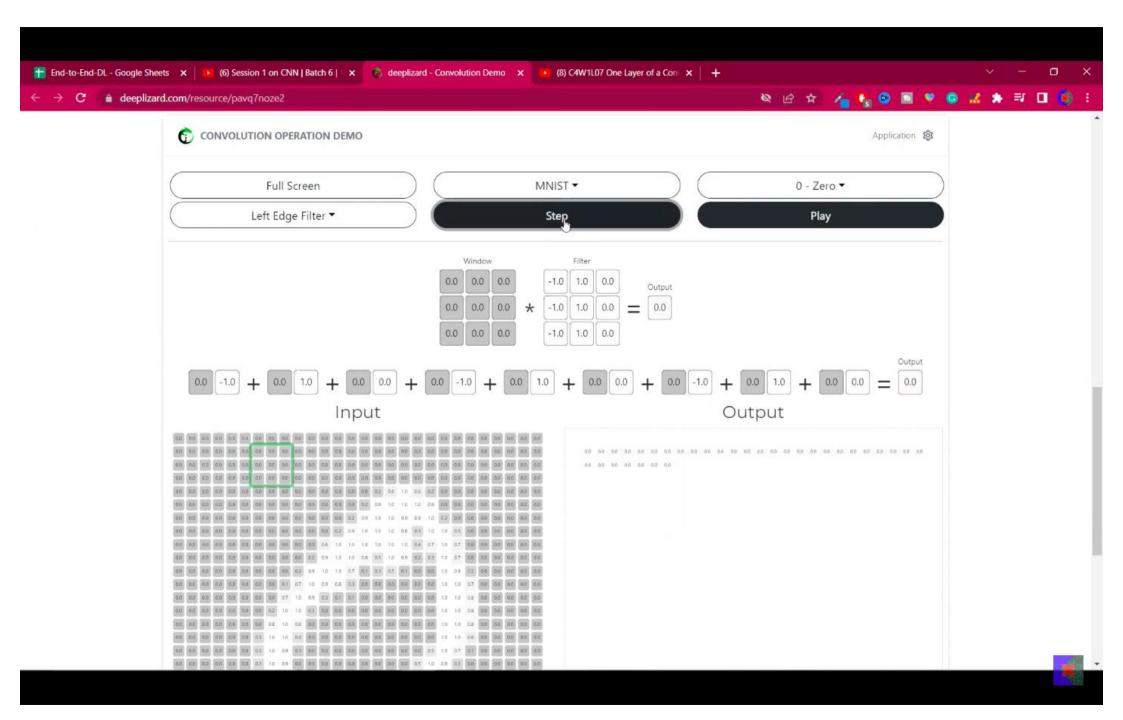


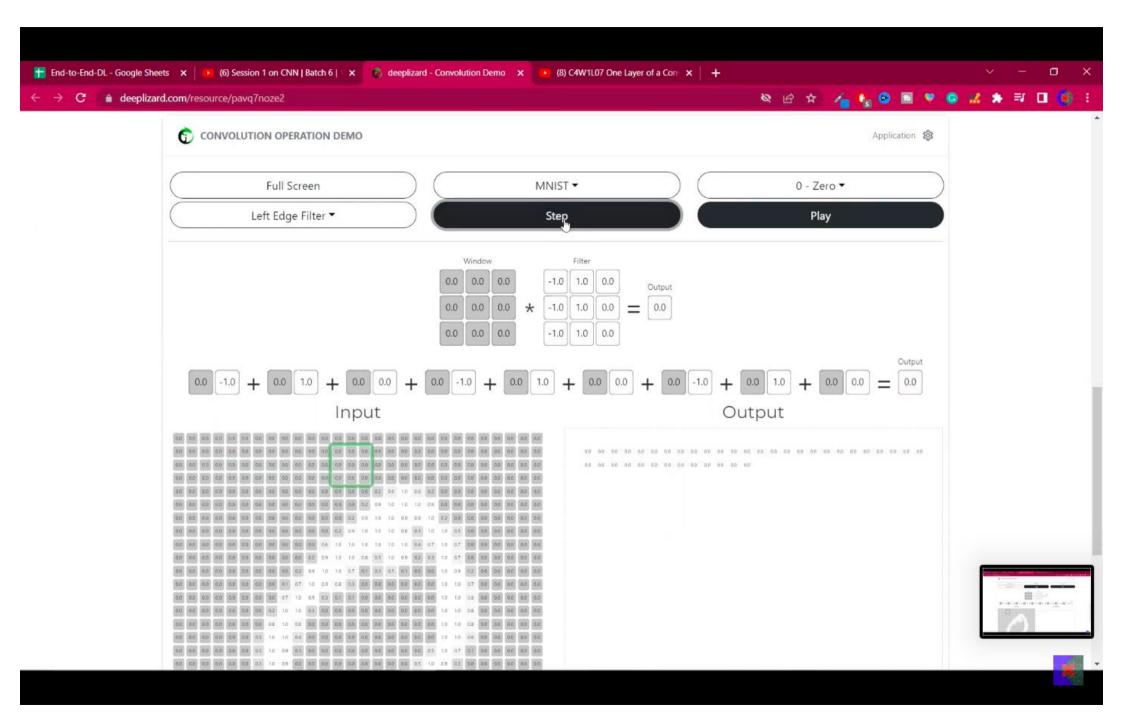
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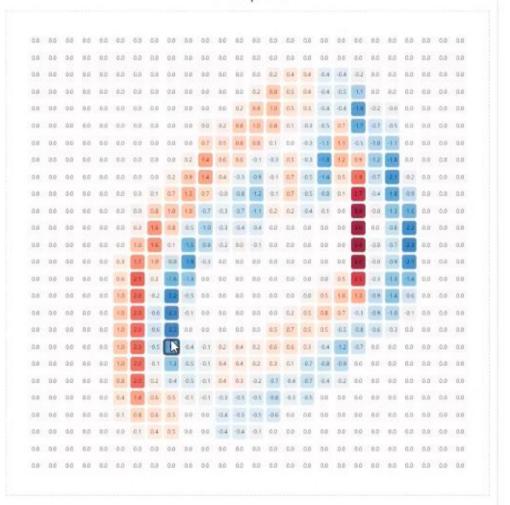






Input

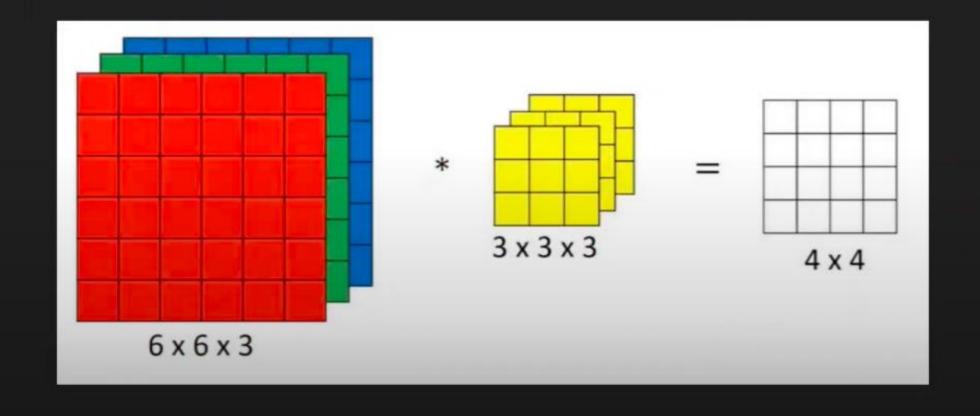
Output

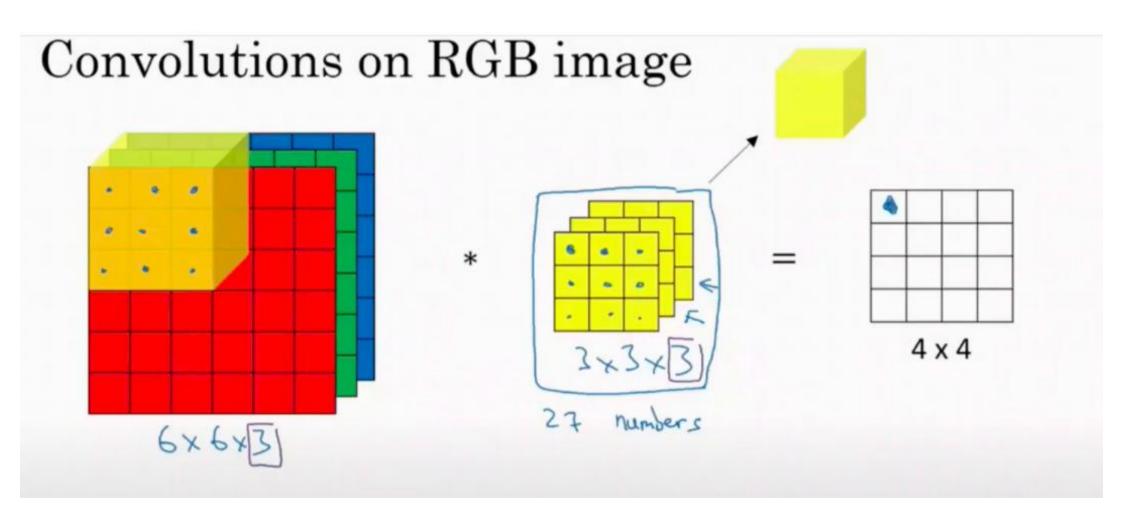


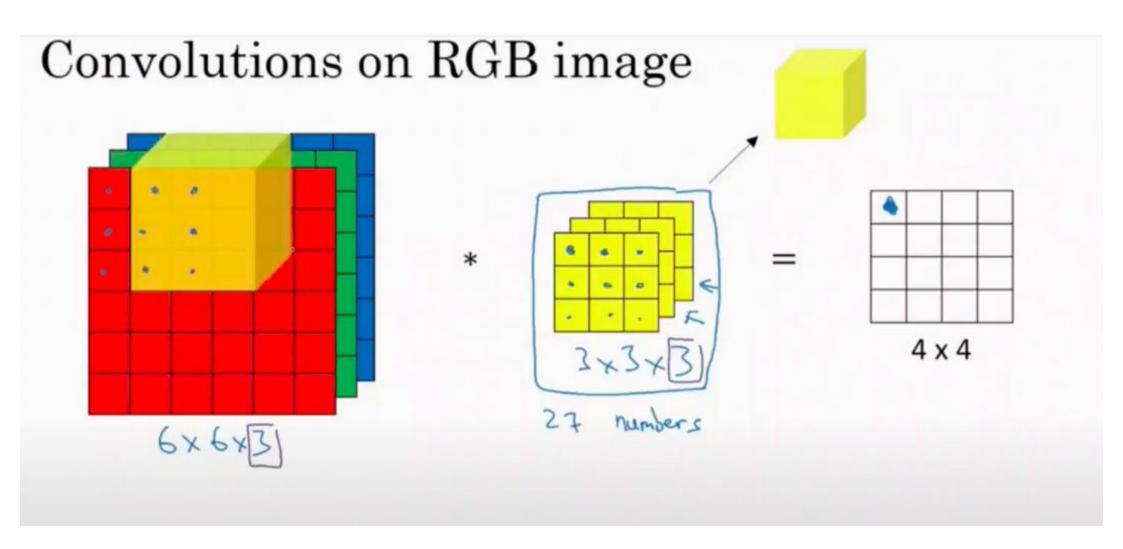
Working with RGB Images

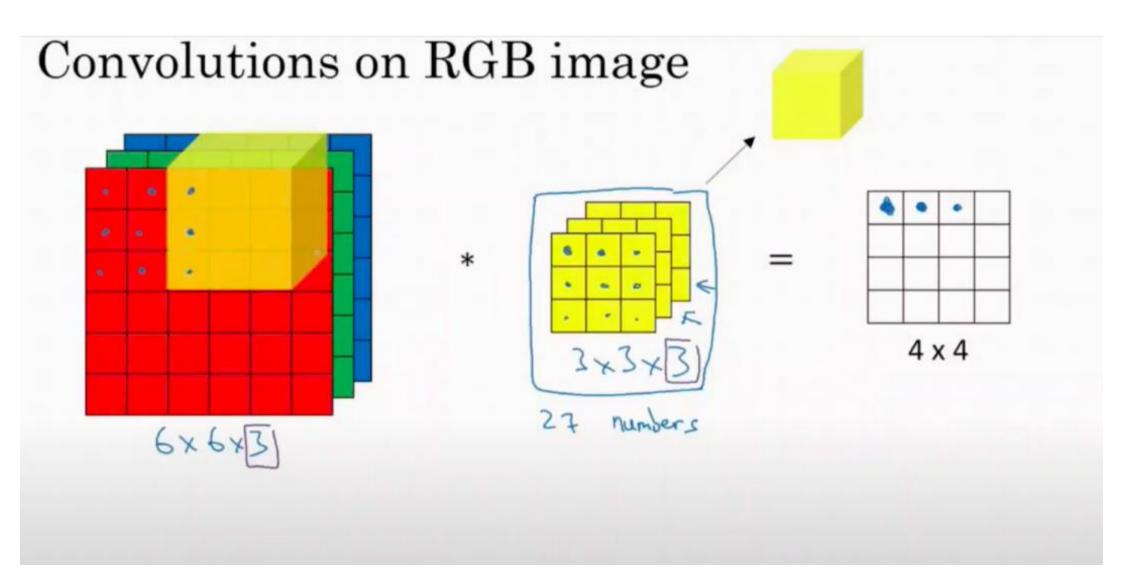
19 August 2022

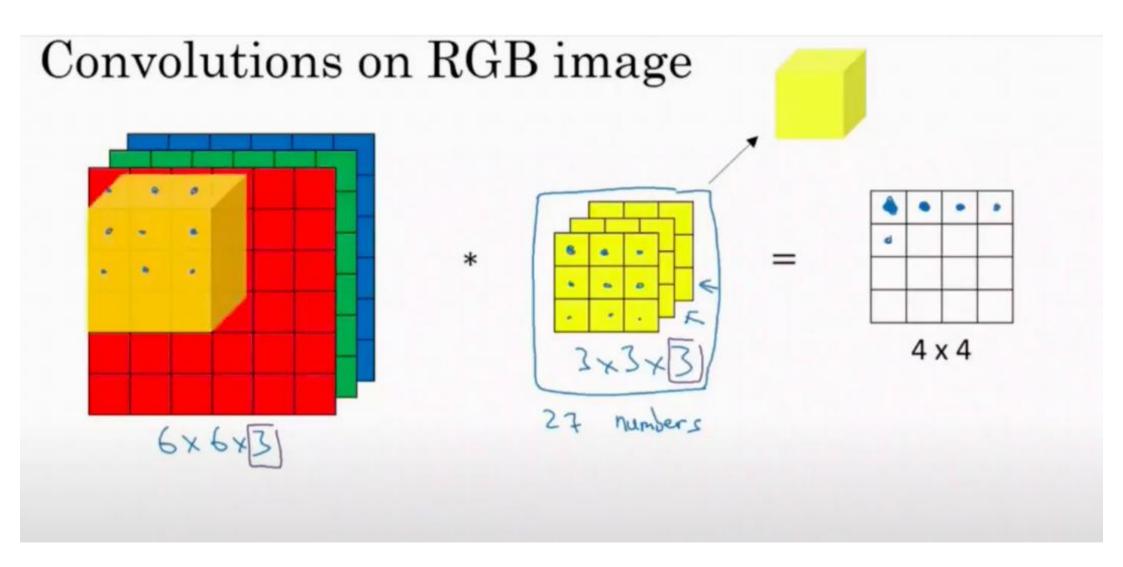
16:54

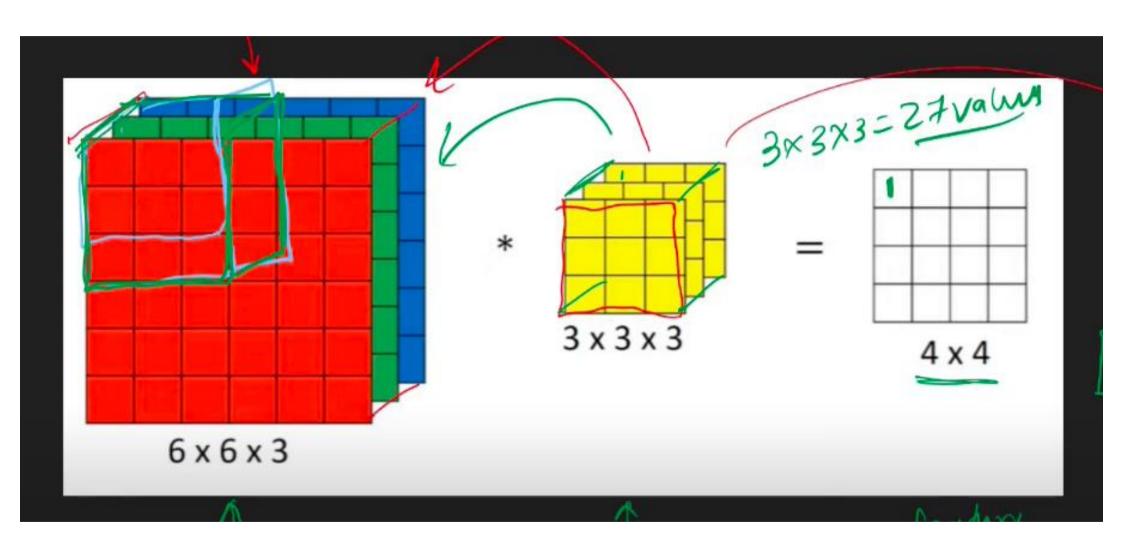












Multiple Filters

23 August 2022

08:24

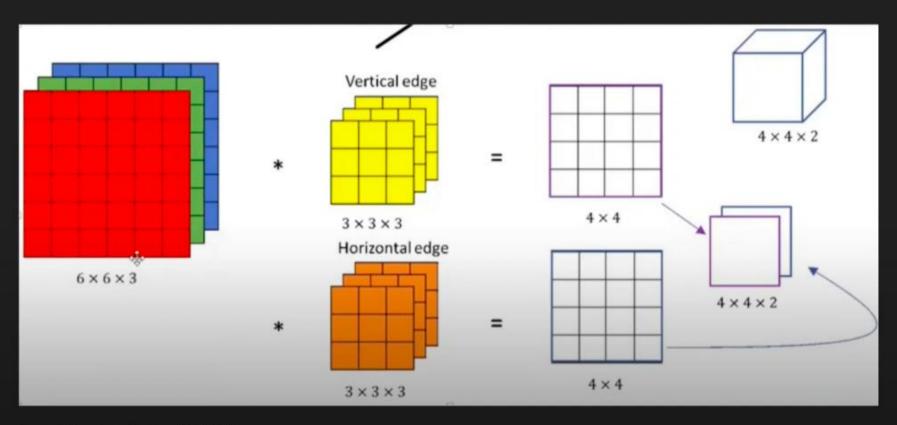


Image taken from Andrew NG's lecture

7	2	3	3	8				
4	4 5	5 3 8	5	5 3 8		3 8	3 8	
3	3	2	8	4				
2	8	7	2	7				
5	4	4	5	4				

1	0	-1
1	0	-1
1	0	-1

7x1+4x1+3x1+ 2x0+5x0+3x0+ 3x-1+3x-1+2x-1 = 6

6	

7	2	3	3	8
4	5	3	8	4
3	3	2	8	4
2	8	7	2	7
5	4	4	5	4

1	0	-1
1	0	-1
1	0	-1

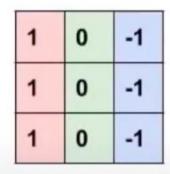
6	-9	-8
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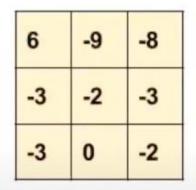
7	2	3	3	8
4	5	3	8	4
3	3	2	8	4
2	8	7	2	7
5	4	4	5	4

1	0	-1
1	0	-1
1	0	-1

6	-9	-8
-3	-2	-3
-3		

7	2	3	3	8
4	5	3	8	4
3	3	2	8	4
2	8	7	2	7
5	4	4	5	4





What is Padding?

26 August 2022

14:26

7	2	3	3	8
4	5	3	8	4
3	3	2	8	4
2	8	7	2	7
5	4	4	5	4

	1	0	-1		6	
	1	0	-1	=		
	1	0	-1			
1	2x0-	+5x0-	+3x1+ +3x0+ 1+2x-1			

$$f \times f \rightarrow (n-f+1)(n-f+1)$$

 $3 \times 3 \qquad (5-3+1) = 3 \times 3$

0	0	0	0	0	0	0
0	60	113	56	139	85	0
0	73	121	54	84	128	0
0	131	99	70	129	127	0
0	80	57	115	69	134	0
0	104	126	123	95	130	0
0	0	0	0	0	0	0

0	-1	0
-1	5	-1
0	-1	0

114	328	-26	

			- P			
0	0	0	0	0	0	0
0	60	113	56	139	85	0
0	73	121	54	84	128	0
0	131	99	70	129	127	0
0	80	57	115	69	134	0
0	104	126	123	95	130	0
0	0	0	0	0	0	0

0	-1	0
-1	5	-1
0	-1	0

114	328	-26	470	158
53	266	-61	-30	

0	0	0	0	0	0	0
0	60	113	56	139	85	0
0	73	121	54	84	128	0
0	131	99	70	129	127	0
0	80	57	115	69	134	0
0	104	126	123	95	130	0
0	0	0	0	0	0	0

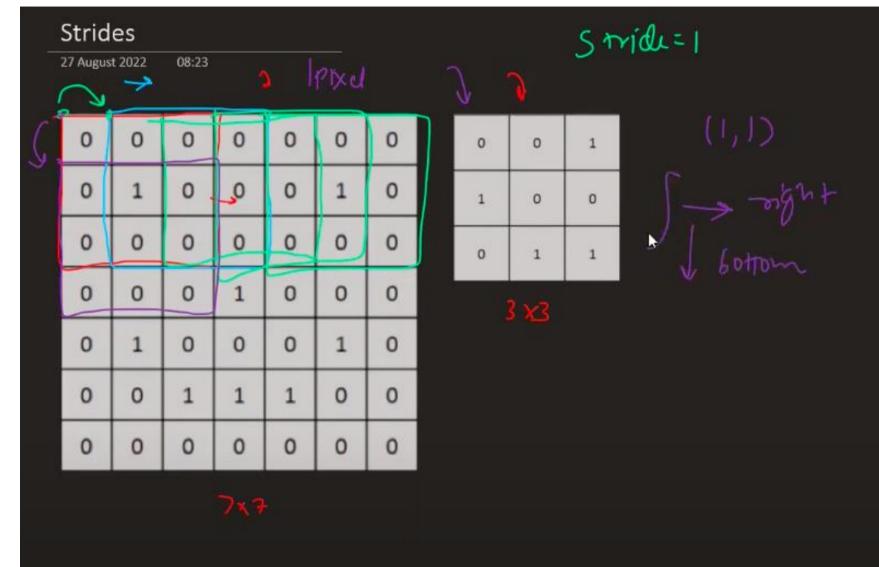
0	-1	0
-1	5	-1
0	-1	0

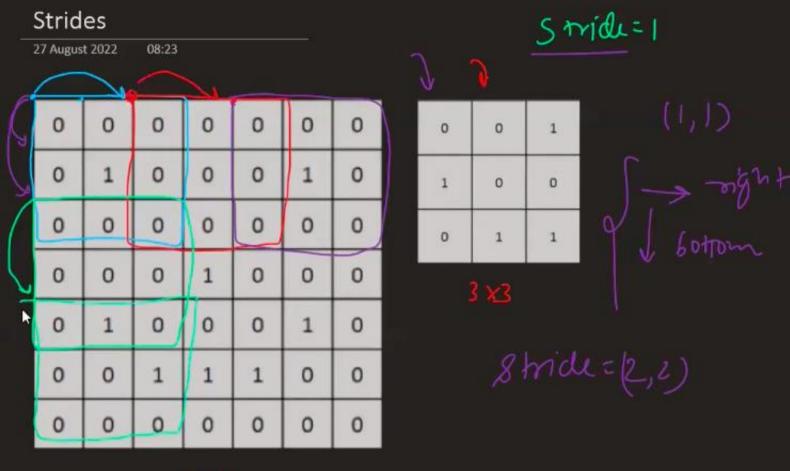
114	328	-26	470	158
53	266	-61	-30	344
403	116	-47	295	244

0	0	0	0	0	0	0
0	60	113	56	139	85	0
0	73	121	54	84	128	0
0	131	99	70	129	127	0
0	80	57	115	69	134	0
0	104	126	123	95	130	0
0	0	0	0	0	0	0

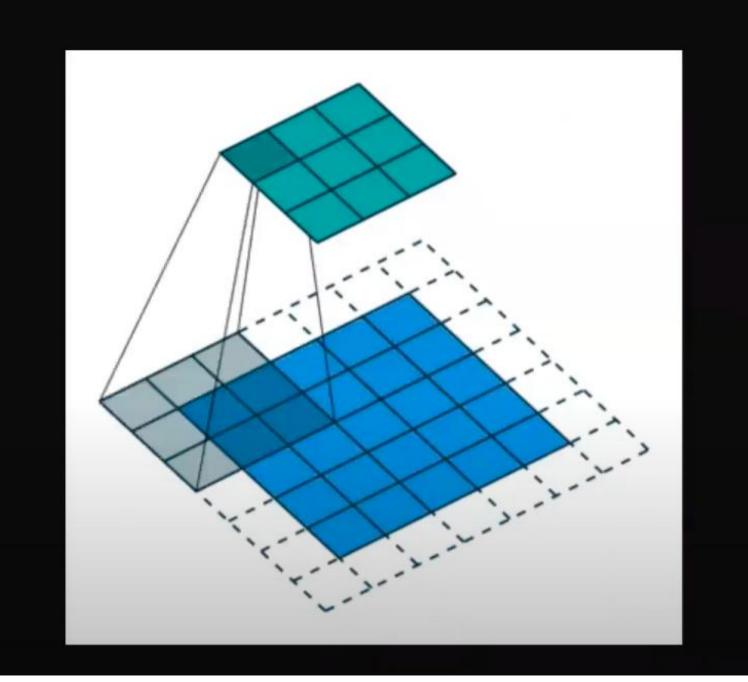
0	-1	0
-1	5	-1
0	-1	0

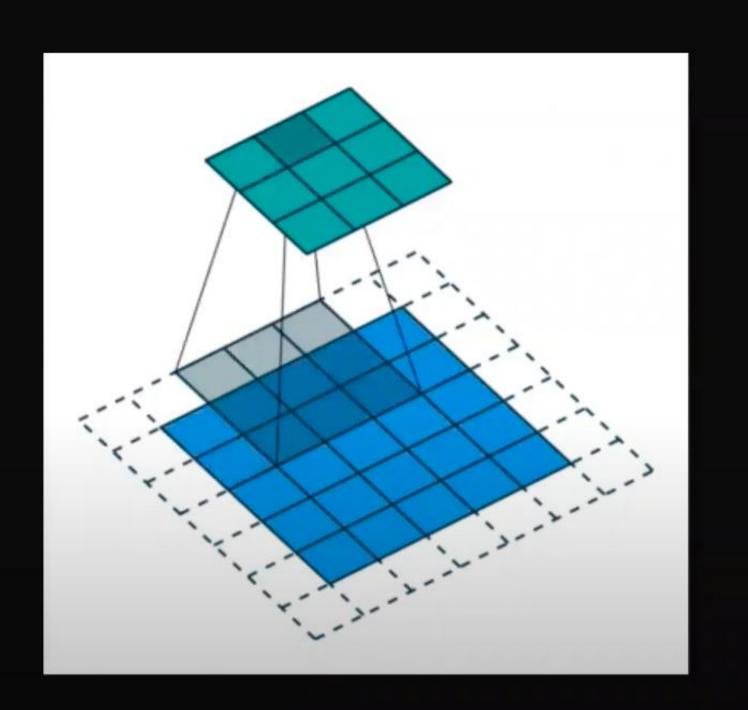
114	328	-26	470	158
53	266	-61	-30	344
403	116	-47	295	244
108	-135	256	-128	344
314	346	279	153	421

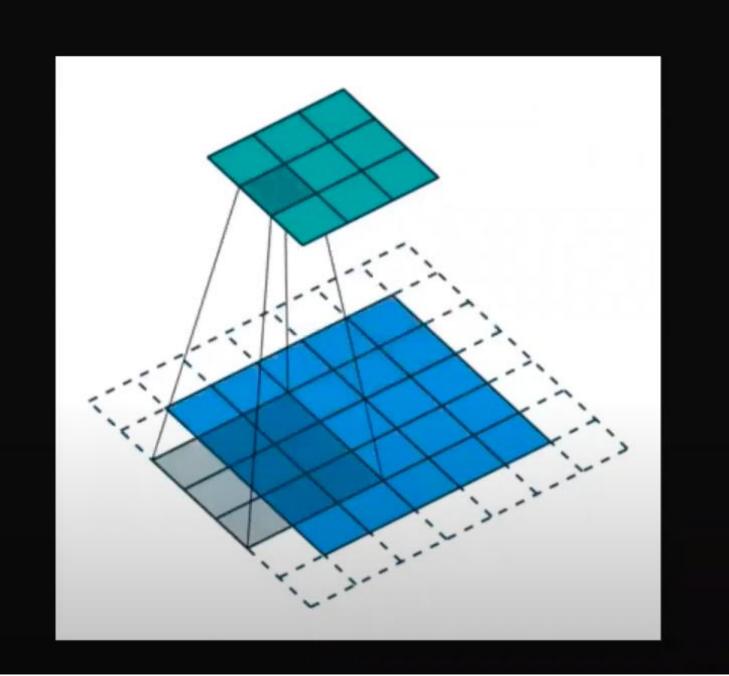




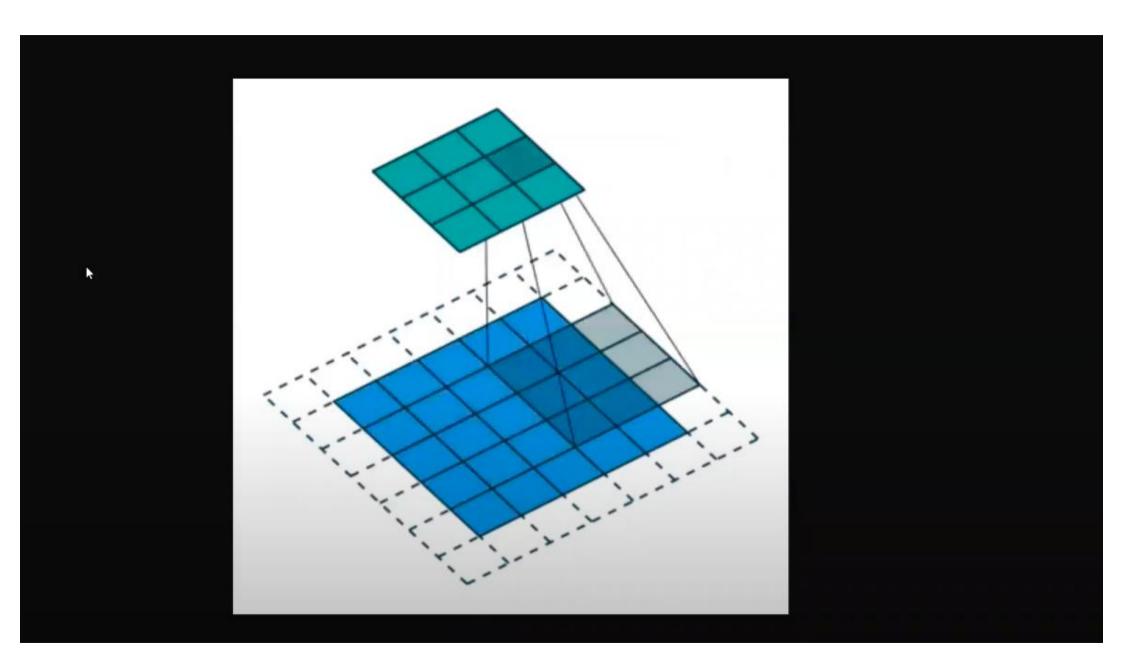
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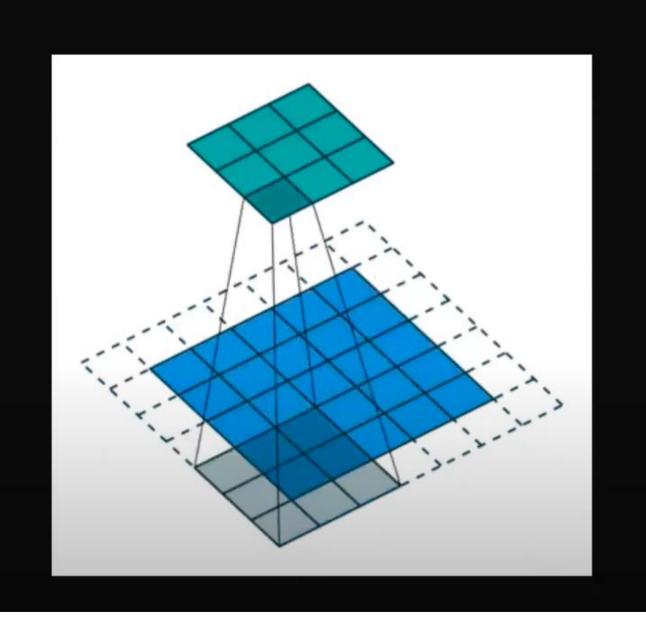


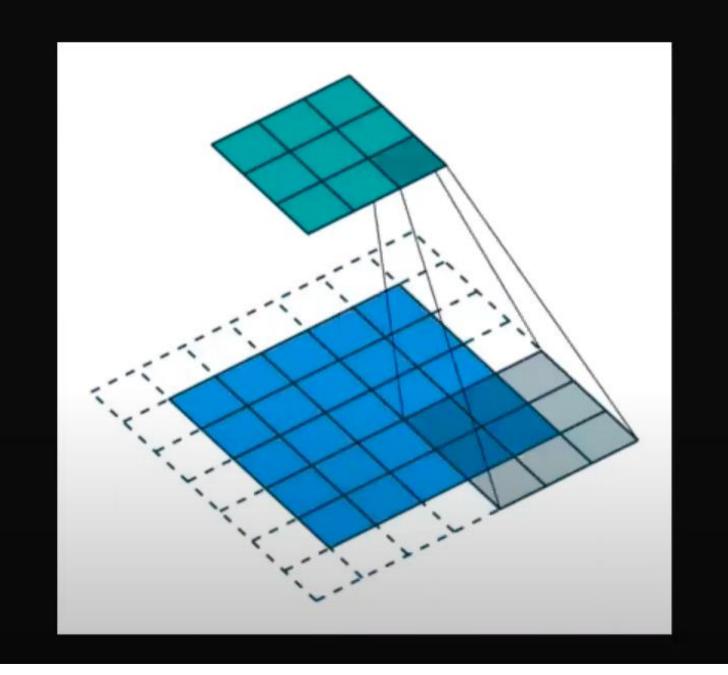


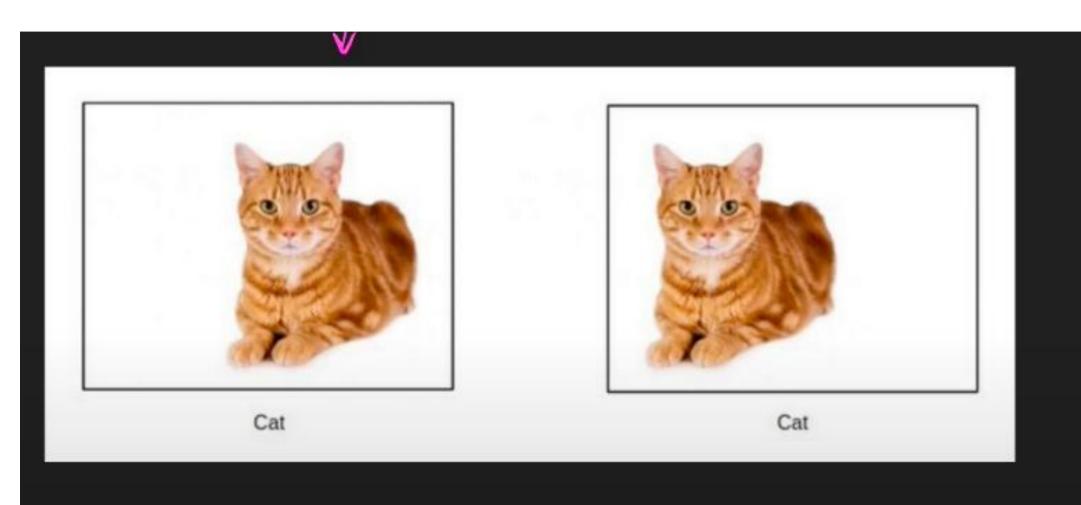


k









Pooling

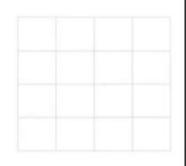
01 September 2022

09:55

0	0	0	0	0	0
0	0	0	0,	0	0
0	0	0	0	0	0
255	255	255	255	255	255
255	255	255	255	255	255
255	255	255	255	255	255

*

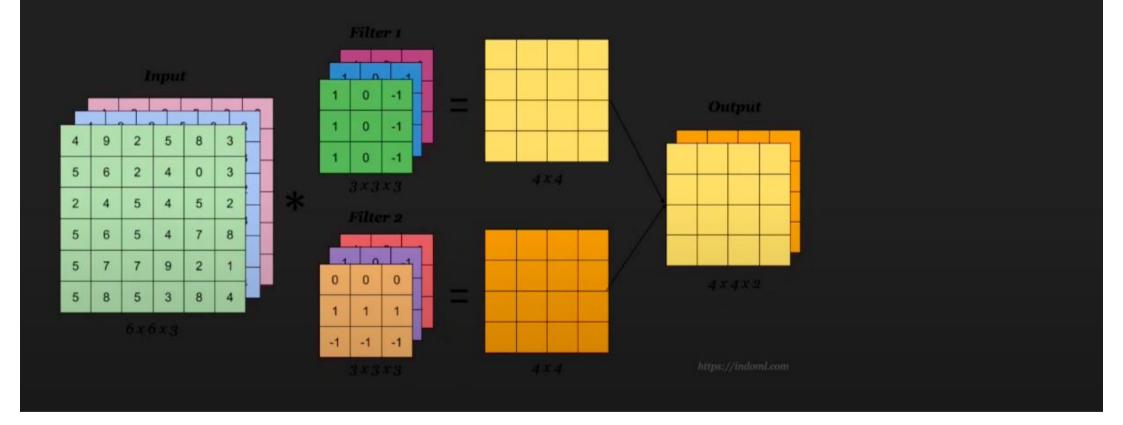
=

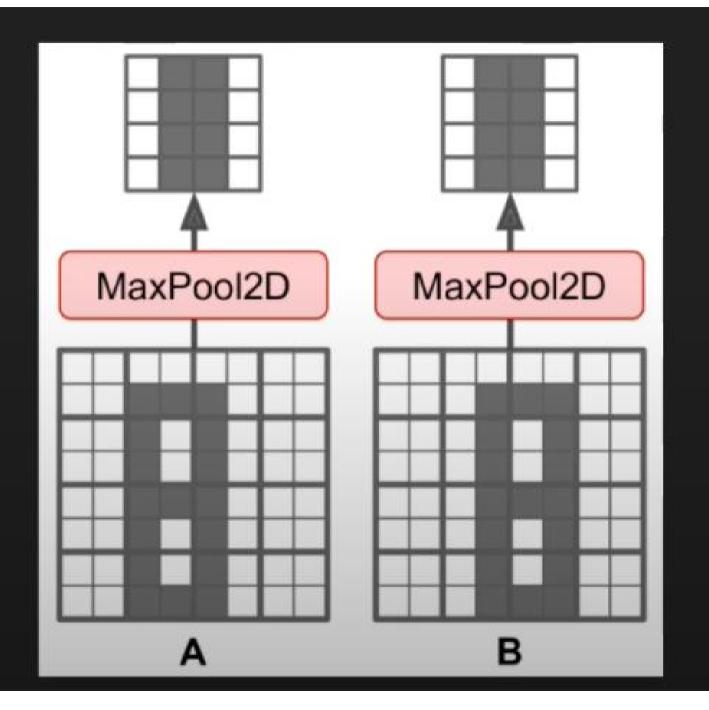


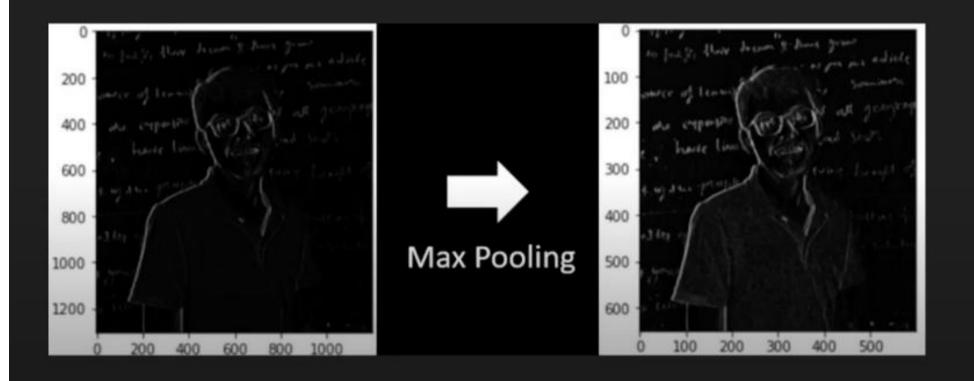


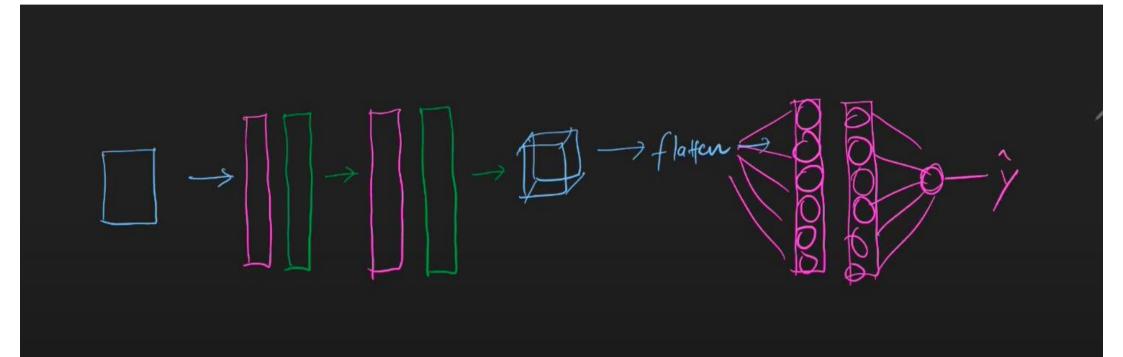
01 September 2022

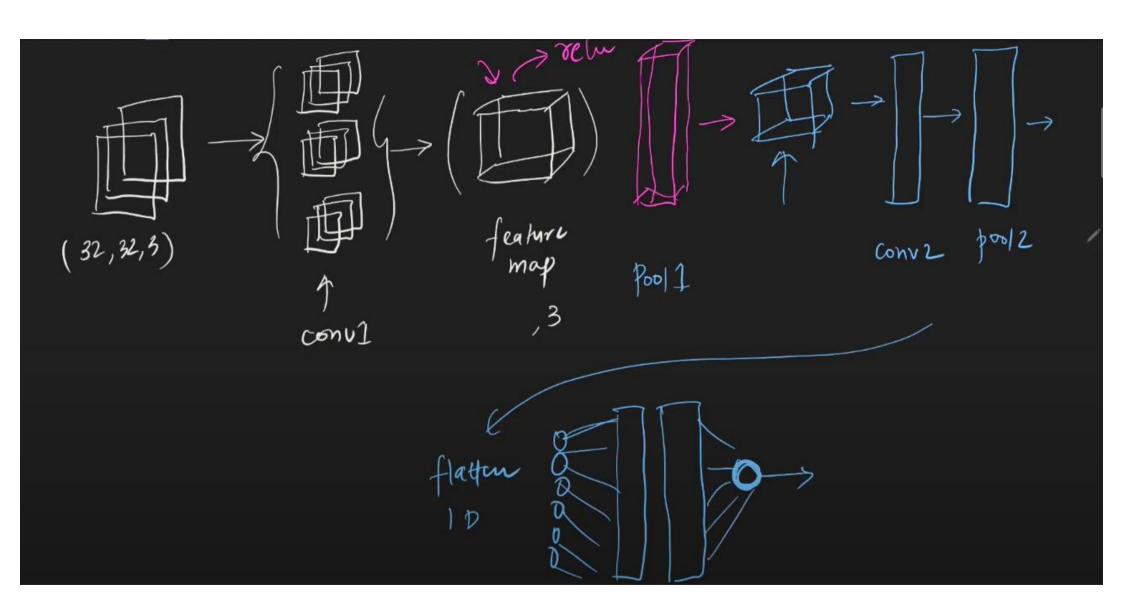
09:56

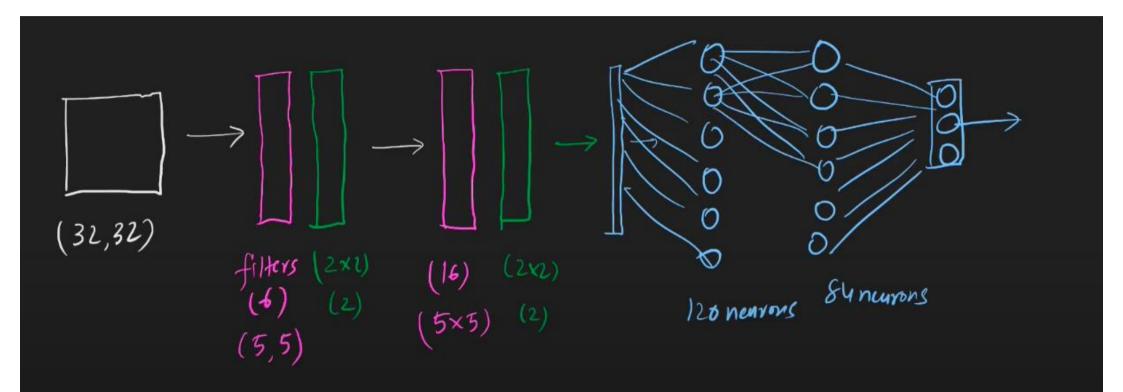


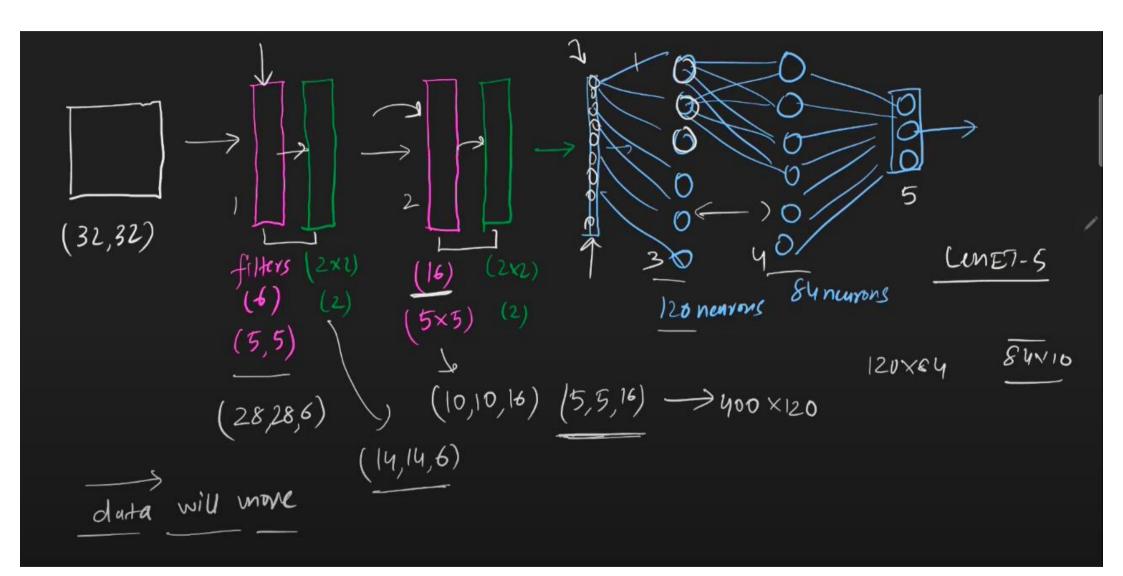












```
import tensorflow
     from tensorflow import keras
     from keras.layers import Dense, Conv2D, Flatten, AveragePooling2D
     from keras import Sequential
     from keras.datasets import mnist[
[ ] (X_train, y_train), (X_test, y_test) = mnist.load_data()
     Downloading data from <a href="https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz">https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz</a>
     11501568/11490434 [============== ] - 0s Ous/step
LeNet Architecture
```

LeNet Architecture

```
model = Sequential()
model.add(Conv2D(6,kernel_size=(5,5),padding='valid', activation='tanh', input_shape=(32,32,1)))
model.add(AveragePooling2D(pool_size=(2, 2), strides=2, padding='valid'))
model.add(Conv2D(16,kernel_size=(5,5),padding='valid', activation='tanh'))
model.add(AveragePooling2D(pool_size=(2, 2), strides=2, padding='valid'))
model.add(Flatten())
model.add(Dense(120, activation='tanh'))
model.add(Dense(84, activation='tanh'))
model.add(Dense(10, activation='softmax'))
```

moder: Sequential_I	Model:	"sequential	1
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Layer (type)	Output Shape	Param # ======
conv2d_2 (Conv2D)	(None, 28, 28, 6)	156
<pre>average_pooling2d_2 (Averag ePooling2D)</pre>	(None, 14, 14, 6) I	0
conv2d_3 (Conv2D)	(None, 10, 10, 16)	2416
<pre>average_pooling2d_3 (Averag ePooling2D)</pre>	(None, 5, 5, 16)	0
flatten_1 (Flatten)	(None, 400)	0
dense_3 (Dense)	(None, 120)	48120
dense_4 (Dense)	(None, 84)	10164
dense_5 (Dense)	(None, 10)	850



