Convolution Neural Network

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Hyper Parameter Calculate Formal's

Convolutional Layer Size Calculate Formula

$$K = \frac{(W - F + 2P)}{S} + 1 \tag{1}$$

Pooling Layer Size Calulate Formula

$$K = \frac{(W-F)}{S} + 1 \tag{2}$$

- W: Input volume size
- F: Filter Size
- S: Stride (Sampling in the Case of Pooling)
- P: Padding



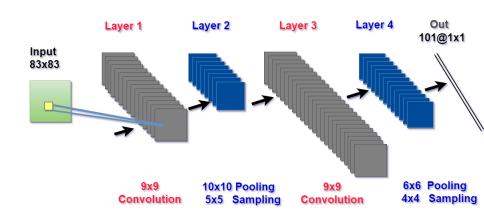


Figure 1: CNN



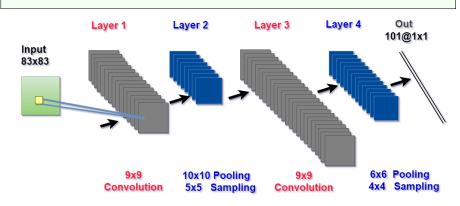


Figure 2: CNN



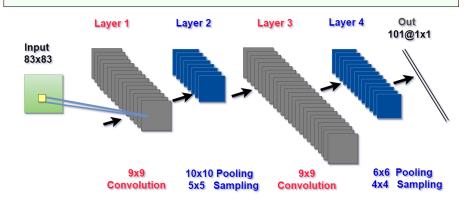


Figure 3: CNN



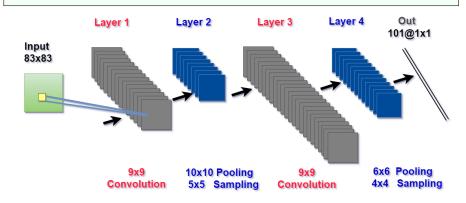


Figure 4: CNN



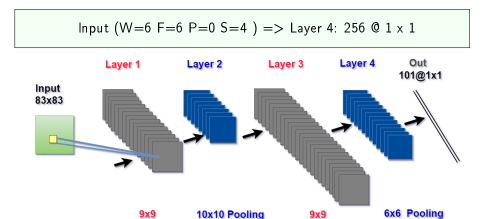


Figure 5: CNN

5x5 Sampling

Convolution



Convolution

4x4 Sampling

Learnable Parameters For CNN

- Input Layer: No parameter learning
- Convolutional Layers:

Number of parameter =
$$(n * m * l + 1) * k$$
 (3)

n and m is filter size I is feature maps as input k is feature maps as outputs

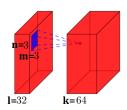


Figure 6: Convolution Layer

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Learnable Parameters For CNN

- Opening layers: No Parameter learning
- Fully-connected layers:

Number of parameter =
$$(n+1) * m$$
 (4)

n is the number inputs m is the numbers outputs

Output layer:

Number of parameter =
$$(n+1) * m$$
 (5)

n is the number inputs m is the numbers outputs



Architecture of LexNet-5

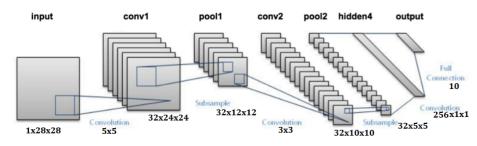


Figure 7: LexNet-5

Learnable Parameters For LexNet-5

Name	Size	Formula	Parameters
0 Input	1x28x28	-	0
1 Convolution (28-(5-1))=24	32x24x24	$(n \times m \times l + 1) \times k$	(5x5x1+1)x32=832
2 Maxpool	32x12x12	-	0
3 Convolution (12-(3-1))=10	32×10×10	$(n \times m \times l + 1) \times k$	(3x3x32+1)x32=9248
4 Maxpool	32x5x5	-	0
5 Fully Con- nected	256	(n+1)×m	(32x5x5+1)x256=205056
6 Output	10	$(n+1)\times m$	$(256+1)\times10=2570$

Table 1: Learnable Parameters For LexNet-5

AlexNet Architecture

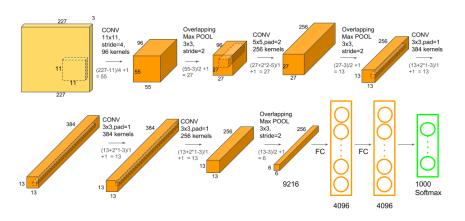


Figure 8: AlexNet Architecture

AlexNet Architecture

- AlexNet consists of 5 Convolutional Layers and 3 Fully Connected Layers.
- The first two Convolutional layers are followed by the Overlapping Max Pooling layers.
- The third, fourth and fifth convolutional layers are connected directly the output of which goes into a series of two fully connected layers.
- The second fully connected layer feeds into a softmax classifier with 1000 class labels.

Size / Operation	Filter	Depth	Stride	Padding	Number of Parameter
3* 227 * 227					
Conv1	11 * 11	96	4		(11*11*3 + 1) * 96=34944
96 * 55 * 55					
Max Pooling	3*3		2		
96 * 27 * 27					
Conv2	5*5	256	1	2	(5 * 5 * 96 + 1) * 256=614656
256 * 27 * 27					
Max Pooling	3*3		2		
256 * 13 * 13					
Conv3	3*3	384	1	1	(3 * 3 * 256 + 1) * 384=885120
384 * 13 * 13					
Conv4	3*3	384	1	1	(3 * 3 * 384 + 1) * 384=1327488
384 * 13 * 13					
Conv5	3*3	256	1	1	(3 * 3 * 384 + 1) * 256=884992
256 * 13 * 13					
Max Pooling	3*3		2		
256 * 6 * 6					
FC6					256 * 6 * 6 * 4096=37748736
4096					
FC7					4096 * 4096=16777216
4096					
FC8					4096 * 1000=4096000
1000 classes					
Overall					62369152=62.3 million
Conv VS FC	Conv:3.7million (6%), FC: 58.6 million (94%)				

Figure 9: Learnable Parameters For AlexNet

VGG Architecture



Figure 10: Architecture of VGG.

Feature Extractor and Fine-tune

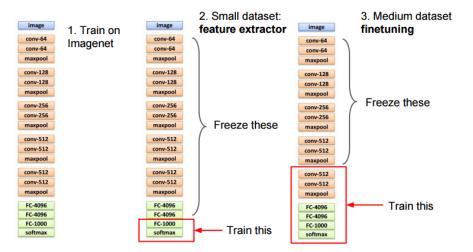


Figure 11: Feature Extractor and Fine-tune

- www.jefkine.com
- 2 www.learnopencv.com