

# Parameter Calculation of Neural Network

CSE 4237  
Soft Computing

# Model Architecture

```
NeuralNetwork(  
    (cnn_layer_1): Conv2d(1, 16, kernel_size=(5, 5), stride=(1, 1), padding=(2, 2))  
    (cnn_layer_2): Conv2d(16, 32, kernel_size=(5, 5), stride=(1, 1), padding=(2, 2))  
    (flatten): Flatten(start_dim=1, end_dim=-1)  
    (maxpool): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil_mode=False)  
    (linear_layer_1): Linear(in_features=1568, out_features=512, bias=True)  
    (linear_layer_2): Linear(in_features=512, out_features=128, bias=True)  
    (linear_layer_3): Linear(in_features=128, out_features=10, bias=True)  
    (relu): ReLU()  
    (sigmoid): Sigmoid()  
    (dropout): Dropout(p=0.2, inplace=False)  
)
```

```
def forward(self, x):  
    x = self.cnn_layer_1(x)  
    x = self.dropout(x)  
    x = self.relu(x)  
    x = self.maxpool(x)  
    x = self.cnn_layer_2(x)  
    x = self.dropout(x)  
    x = self.relu(x)  
    x = self.maxpool(x)  
    x = self.flatten(x)  
    x = self.linear_layer_1(x)  
    x = self.dropout(x)  
    x = self.relu(x)  
    x = self.linear_layer_2(x)  
    x = self.dropout(x)  
    x = self.relu(x)  
  
    x = self.linear_layer_3(x)  
    return x
```

Number of parameters in a CONV layer would be :

$$((m * n * d) + 1) * k$$

added 1 because of the bias term for each filter.

The same expression can be written as follows:

**((shape of width of the filter \* shape of height of the filter \* number of filters in the previous layer + 1) \* number of filters).**

## Calculation of CNN Output Shape

$$\text{Floor}\left(\frac{(W-F+2P)}{S} + 1\right)$$

# Model Summary

Layer (type)	Output Shape	Param #
Conv2d-1	[-1, 16, 28, 28]	416
Dropout-2	[-1, 16, 28, 28]	0
ReLU-3	[-1, 16, 28, 28]	0
MaxPool2d-4	[-1, 16, 14, 14]	0
Conv2d-5	[-1, 32, 14, 14]	12,832
Dropout-6	[-1, 32, 14, 14]	0
ReLU-7	[-1, 32, 14, 14]	0
MaxPool2d-8	[-1, 32, 7, 7]	0
Flatten-9	[-1, 1568]	0
Linear-10	[-1, 512]	803,328
Dropout-11	[-1, 512]	0
ReLU-12	[-1, 512]	0
Linear-13	[-1, 128]	65,664
Dropout-14	[-1, 128]	0
ReLU-15	[-1, 128]	0
Linear-16	[-1, 10]	1,290

# Model Summary

```
=====
Total params: 883,530
Trainable params: 883,530
Non-trainable params: 0
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```

# Example

Calculate the number of parameters for the following architecture for the input shape of 1024.

input→FCwB(256)→Dropout(0.1)→FC(128)→tanh→Maxpool(3)→FC(3)→output

Here, FCwB(x) denotes a Fully Connected Layer of x neurons with Bias,

FC(x) denotes a Fully Connected Layer of x neurons without Bias,

Dropout (y) denotes a Dropout Layer that drops input at a rate of y, and

Maxpool(x) denotes a flat Maxpooling layer that has both window size and stride size of x.



# Example

Layer	Output Shape	No. of Parameter
input	1024	0
FCwB	256	$(1024 \times 256 + 256) = 2,62,400$
Dropout	256	0
FC	128	$256 \times 128 = 32,768$
tanh	128	0
Maxpool	$\text{floor}((128 - 3 + 2 \times 0) / 3) = 42$	0
FC	3	$42 \times 3 = 126$
Output	3	0
	Total	2,95,294

# Reference

[https://towardsdatascience.com/understanding-and-calculating-the-number-of-parameters-in-convolution-neural-networks-cnns-fc88790d530d#:~:text=Number%20of%20parameters%20in%20a%20CONV%20layer%20would%20be%20%3A%20\(,1\)\\*number%20of%20filters\).](https://towardsdatascience.com/understanding-and-calculating-the-number-of-parameters-in-convolution-neural-networks-cnns-fc88790d530d#:~:text=Number%20of%20parameters%20in%20a%20CONV%20layer%20would%20be%20%3A%20(,1)*number%20of%20filters).)