## DEEP LEARNING FOR COMPUTER VISION

Week6



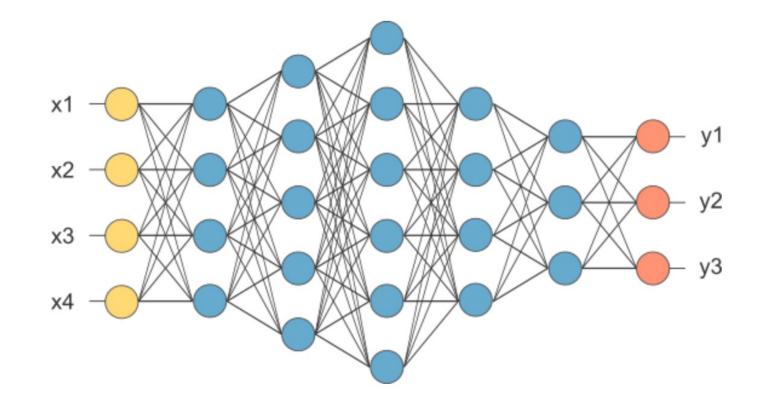
Dr. Tuchsanai. PloySuwan

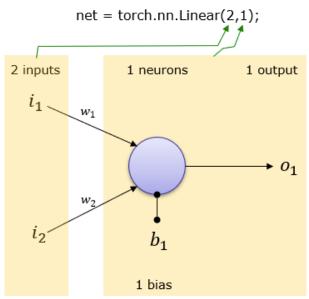


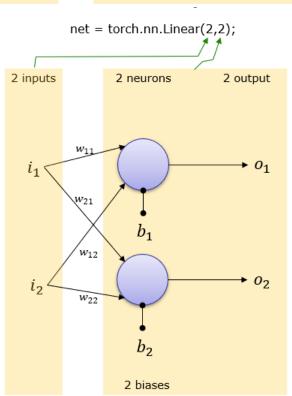
สัปดาห์ที่	หัวข้อ	วิธีการเรียน	ผลการเรียนรู้	หมายเหตุ
1	The basic concept of deep learning and their applications in computer vision	บรรยาย	CLO-1	Chapter 1
2	Introduction to Images and Videos Basics with Python and OpenCV	บรรยาย	CLO-1	Chapter 1
3	Advance in OpenCV	บรรยาย	CLO-1	Chapter 1
4	Introduction to Pytorch	บรรยาย	CLO-2,	Chapter 2
			CLO-3,	
			CLO-4	
5	Fundamental Neural network	บรรยาย	CLO-2	Chapter 2
6	Stochastic optimization methods	บรรยาย	CLO-4	Chapter 2
7	Convolutional neural network for image classification	บรรยาย	CLO-2, CLO-3	Chapter 3
9	Modern architectures such as VGG, ResNet, and Efficientnet	บรรยาย	CLO-3,	Chapter 3
10	Transfer learning and fine tuning	บรรยาย	CLO-5	Chapter 3
11	Deep learning for Segmentation for images	บรรยาย	CLO-6	Chapter 4
12	Deep learning Object detection and Object Tracking	บรรยาย	CLO-7	Chapter 4
13	GAN architecture (Generative Adversarial Network) for photo-realistic images	บรรยาย	CLO-8	Chapter 5
14	Generative image from text CLIP model	บรรยาย	CLO-8	Chapter 5
15	Discussions	บรรยาย		

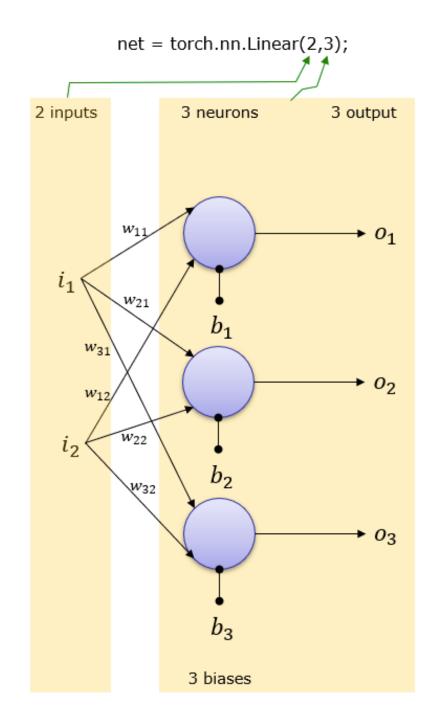
### Fully Connected Layer

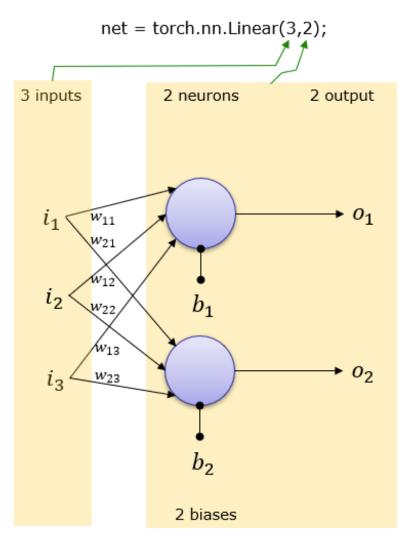
The layer we call as FC layer, we flattened our matrix into vector and feed it into a fully connected layer like a neural network.



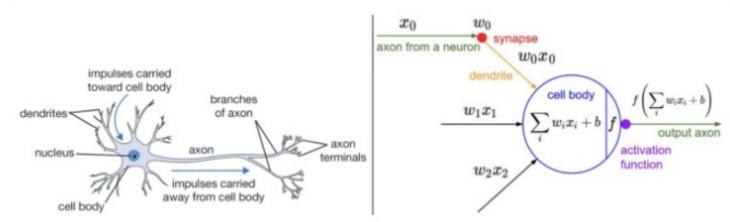








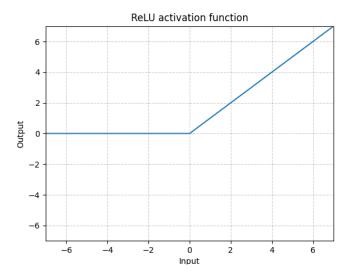
#### **Activation functions in Neural Networks**



A cartoon drawing of a biological neuron (left) and its mathematical model (right).

$$\operatorname{ReLU}(x) = (x)^+ = \max(0, x)$$

$$ext{LeakyRELU}(x) = egin{cases} x, & ext{if } x \geq 0 \\ ext{negative\_slope} imes x, & ext{otherwise} \end{cases}$$



Sigmoid
$$(x) = \sigma(x) = \frac{1}{1 + \exp(-x)}$$

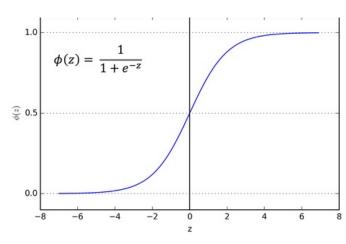


Fig: Sigmoid Function

torch.nn.ReLU() -

torch.nn.Sigmoid

# **Building Neural Network Using PyTorch** Input Layer ∈ R16 Hidden Layer ∈ R12 Hidden Layer ∈ R10 Output Layer ∈ R1

So this is a Fully Connected 16x12x10x1 Neural Network -

- relu activations in hidden layers,
- sigmoid activation in output layer.

Fully Connected (Feed Forward) Network

#### Three Ways to Build a Neural Network in PyTorch

```
import torch
import torch.nn.functional as F
from torch import nn
# define the network class
class MyNetwork(nn.Module):
    def __init__(self):
        # call constructor from superclass
        super().__init__()
        # define network layers
        self.fc1 = nn.Linear(16, 12)
        self.fc2 = nn.Linear(12, 10)
        self.fc3 = nn.Linear(10, 1)
    def forward(self, x):
        # define forward pass
        x = F.relu(self.fc1(x))
        x = F.relu(self.fc2(x))
        x = torch.sigmoid(self.fc3(x))
        return x
# instantiate the model
model = MyNetwork()
```

```
from torch import nn
 2
    # define model architecture
    model = nn.Sequential(
 5
        nn.Linear(16, 12),
        nn.ReLU(),
 6
        nn.Linear(12, 10),
        nn.ReLU(),
 8
        nn.Linear(10, 1),
9
        nn.Sigmoid()
10
11 )
12
    # print model architecture
14 print(model)
```

```
from torch import nn
    from collections import OrderedDict
    # define model architecture
    model = nn.Sequential(OrderedDict([
        ('fc1', nn.Linear(16, 12)),
        ('relu1', nn.ReLU()),
        ('fc2', nn.Linear(12, 10)),
        ('relu2', nn.ReLU()),
 9
        ('fc3', nn.Linear(10, 1)),
10
11
        ('sigmoid', nn.Sigmoid())
    ]))
12
13
    # print model architecture
    print(model)
```

```
import torch
    from torch import nn
    class MyNetwork2(nn.Module):
        def __init__(self):
            super().__init__()
 6
            # define the layers
            self.layers = nn.Sequential(
10
                nn.Linear(16, 12),
                nn.ReLU(),
11
                nn.Linear(12, 10),
12
13
                nn.ReLU(),
14
                nn.Linear(10, 1)
15
16
17
        def forward(self, x):
18
            # forward pass
            x = torch.sigmoid(self.layers(x))
19
20
            return x
21
    # instantiate the model
22
23
    model = MyNetwork2()
24
    # print model architecture
    print(model)
26
```