

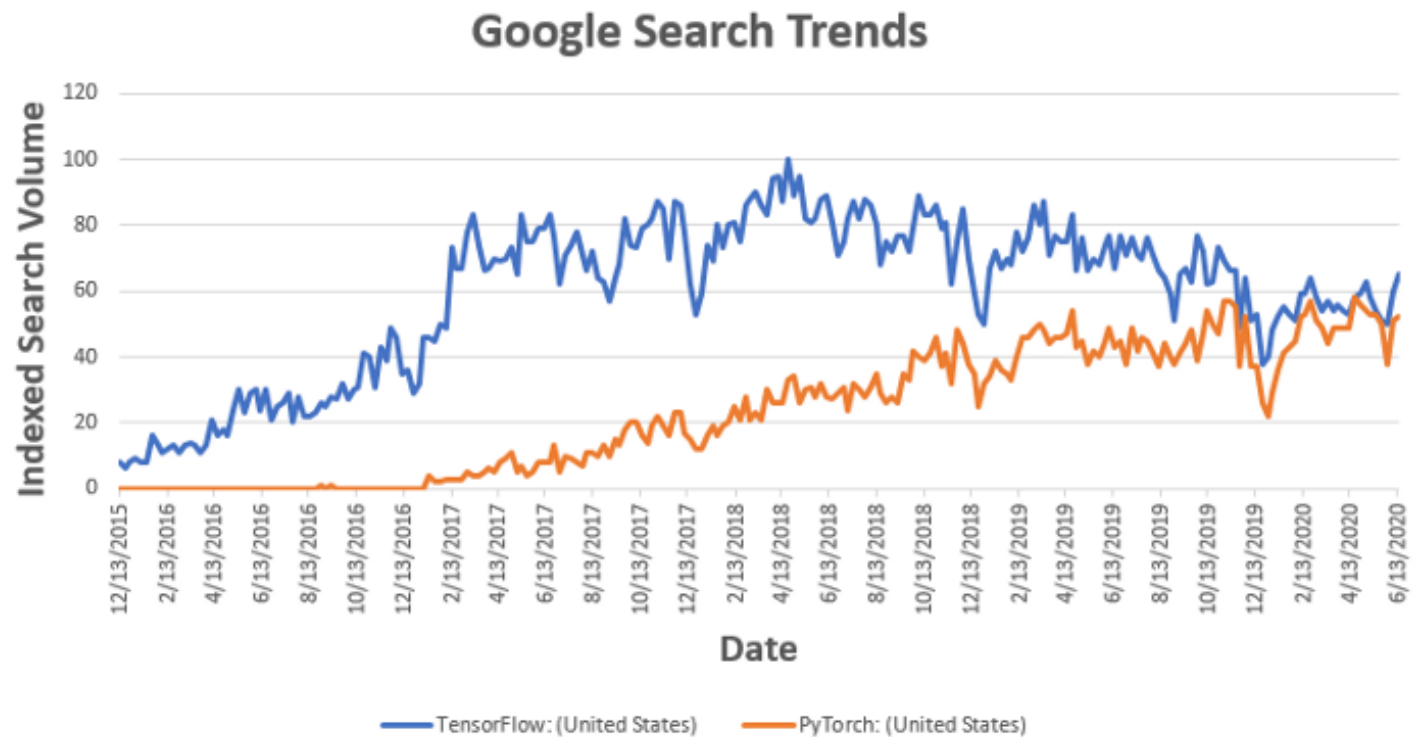
AI Programming

Lecture4 – Pytorch



Pytorch

- Deep learning programming language
 - Tensorflow (Google)
 - Pytorch (FaceBook)
 - Keras

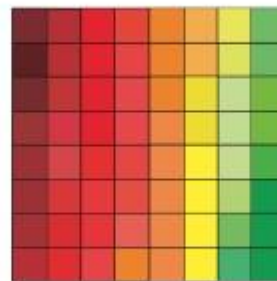


■ What is tensor?

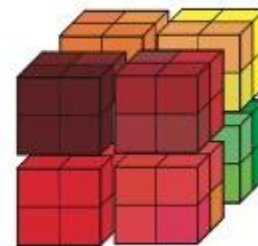
- Tensor is an algebraic object that describes a multilinear relationship between sets of algebraic objects related to a vector space.
- 0-D tensor : scalar value
- 1-D tensor : vector
- 2-D tensor : matrix
- 3-D tensor : cube



1-D
tensor



2-D
tensor



3-D
tensor

Pytorch tensor

- Tensor programming example
 - 2D tensor

1	2	3
4	5	6
7	8	9

```
x = torch.tensor([[1,2,3], [4,5,6], [7,8,9]])  
print(x)
```

```
tensor([[1, 2, 3], [ 4, 5, 6], [7, 8, 9]])
```

```
print("Size:", x.size())  
print("Shape:", x.shape)  
print("랭크(차원):", x.ndimension())
```

```
Size: torch.Size([3, 3])  
Shape: torch.Size([3, 3])  
랭크(차원): 2
```

Pytorch tensor

- Tensor programming example
 - Check the tensor shape

```
import torch

x = torch.tensor([[1, 2, 3], [4, 5, 6], [7, 8, 9]])

print("Size :", x.size())
print("Shape :", x.shape)
print("Dimension (Rank)", x.ndimension())
```

```
Size : torch.Size([3, 3])
Shape : torch.Size([3, 3])
Dimension (Rank) 2

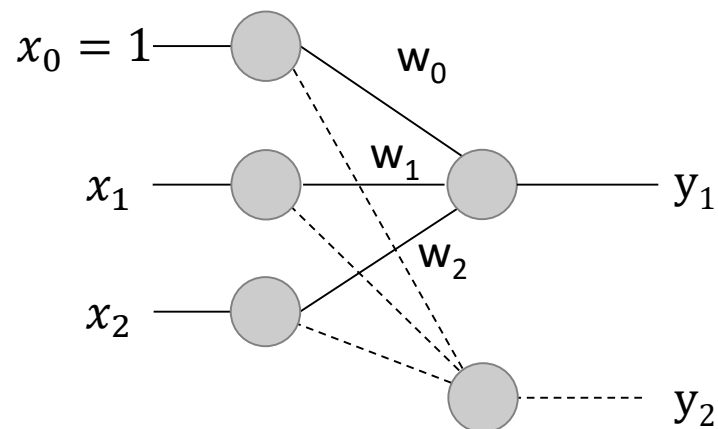
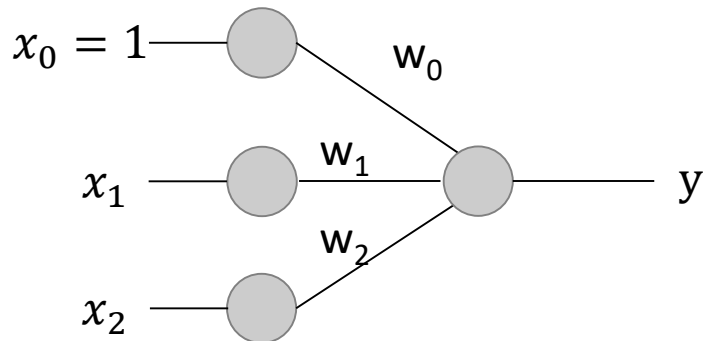
Process finished with exit code 0
```

Pytorch tensor

▪ Matrix multiplication

$$\begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix}_{2 \times 3} \begin{bmatrix} a^* & b^* \\ c^* & d^* \\ e^* & f^* \end{bmatrix}_{3 \times 2} = \begin{bmatrix} aa^*+bc^*+ce^* & ab^*+bd^*+cf^* \\ aa^*+ec^*+fe^* & db^*+ed^*+ff^* \end{bmatrix}_{2 \times 2}$$

▪ Neural Network



Pytorch tensor

▪ Tensor programming example

- matrix w – 5 x 3 random matrix
- matrix x – 3 x 2 constant matrix

```
import torch

w = torch.randn(5,3, dtype=torch.float)
x = torch.tensor([[1.0,2.0], [3.0,4.0], [5.0,6.0]])
wx = torch.mm(w,x) # w의 행은 5, x의 열은 2, 즉 shape는 [5, 2]입니다.
print("wx size:", wx.size())
print("wx:", wx)
```

```
wx size: torch.Size([5, 2])
wx: tensor([[ -3.8183, -4.2929],
           [-4.7751, -5.3661],
           [ 2.1691,  2.2244],
           [-0.0916, -0.8514],
           [-3.8995, -4.3956]])
```

Pytorch tensor

▪ Tensor programming example

- matrix wx – 5 x 3 matrix
- matrix b – 5 x 2 random matrix

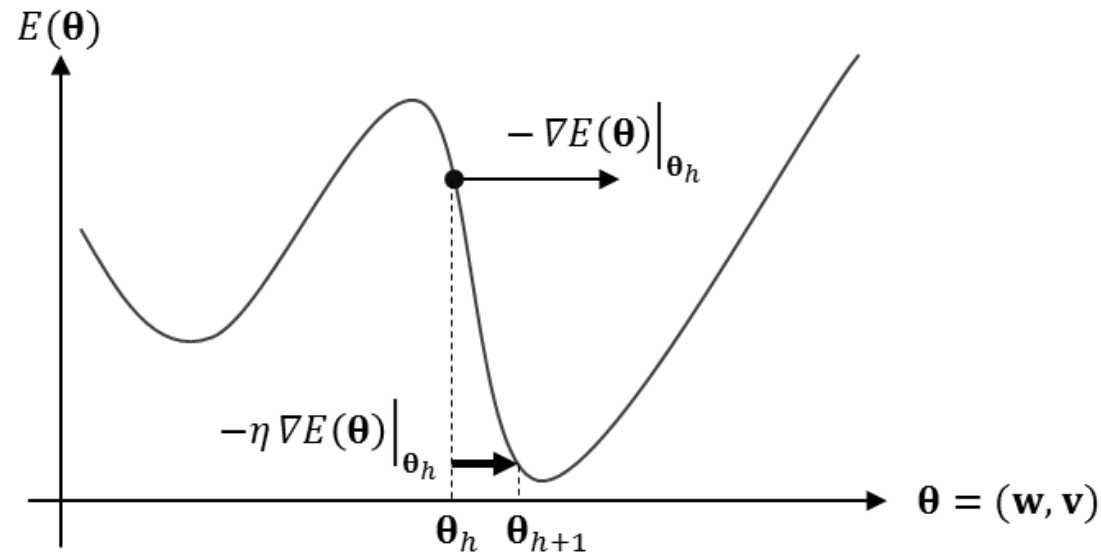
```
b = torch.randn(5,2, dtype=torch.float)

result = wx + b
print("result size:", result.size())
print("result:", result)
```

```
result size: torch.Size([5, 2])
result: tensor([[ -3.7656, -4.9200],
               [-2.7057, -5.4891],
               [ 2.1144,  0.2841],
               [ 1.4777,  0.0384],
               [-3.6158, -3.2231]])
```


Gradient decent

- Auto gradient (Autograd)
 - Minimizing loss using gradient decent
 - Autograd in Pytorch computes the gradient automatically



Gradient decent

- Autograd example

$$l = a^2 = (3w)^2 = 9w^2$$

```
w = torch.tensor(1.0, requires_grad=True)
```

```
a = w*3
```

```
l = a**2
```

“w.grad” contains the gradient of w

```
l.backward()
```

```
print('l을 w로 미분한 값은 {}'.format(w.grad))
```

```
l을 w로 미분한 값은 tensor(18.0)
```

Gradient decent

■ Practice1 – minimization using Autograd

- Find minimum y value

$$y = 0.1x^4 + x^3 + 2x^2 - 5x + 2$$

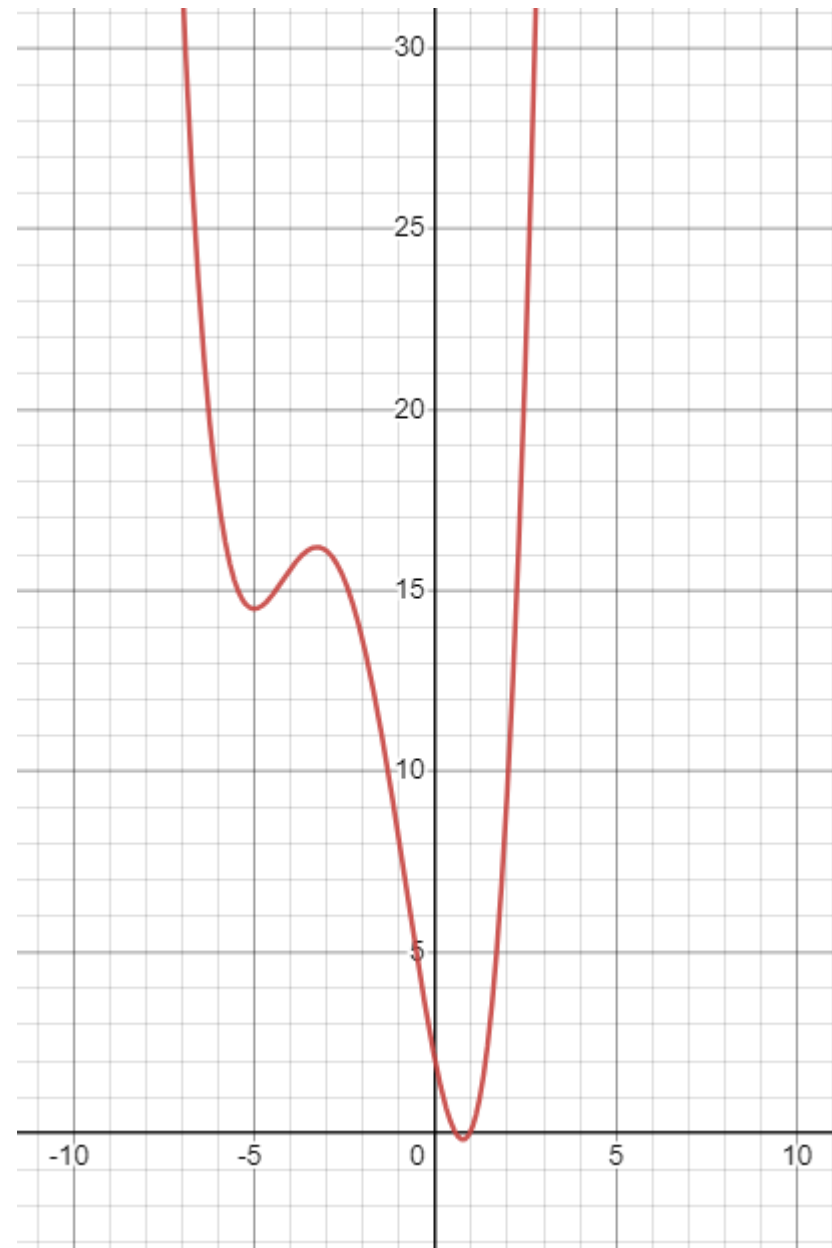
```
import torch

x_t = -10.0
lr = 0.1
for it in range(100):
    x = torch.tensor(x_t, requires_grad=True)
    y = 0.1 * x ** 4 + x ** 3 + 2 * x ** 2 - 5 * x + 2

    print("x : %f  y : %f " % (x_t, y))

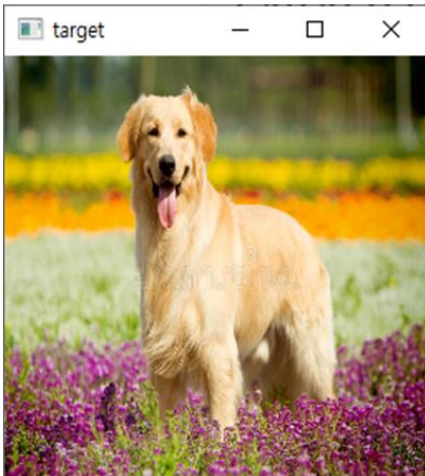
    y.backward()
    # print(format(x.grad))

    x_t = x_t - lr * x.grad
```



Gradient decent

- Practice2 –
random noise to image



```
import torch
import cv2
import numpy as np

size = 256
num_training = 10000
lr = 100

random_tensor = torch.randn((size, size, 3), dtype=torch.float)
random_tensor = torch.clip(random_tensor, min=-1.0, max=1.0)

# ---- random tensor visualization
vis_rt = (random_tensor + 1) / 2 * 255.0
vis_rt = vis_rt.numpy()
# cv2.imshow('rt', vis_rt.astype(np.uint8))
# cv2.waitKey(-1)

# ---- training
target = cv2.imread('./images/dog1.jpg')
target = cv2.resize(target, (size, size))
cv2.imshow('target', target)
# cv2.waitKey(-1)

target = (target / 255.0) * 2 - 1
target = torch.from_numpy(target)

for it in range(num_training):
    random_tensor.requires_grad_(True)
    loss = torch.mean((target - random_tensor)**2)
    loss.backward()

    with torch.no_grad():
        random_tensor = random_tensor - lr*random_tensor.grad

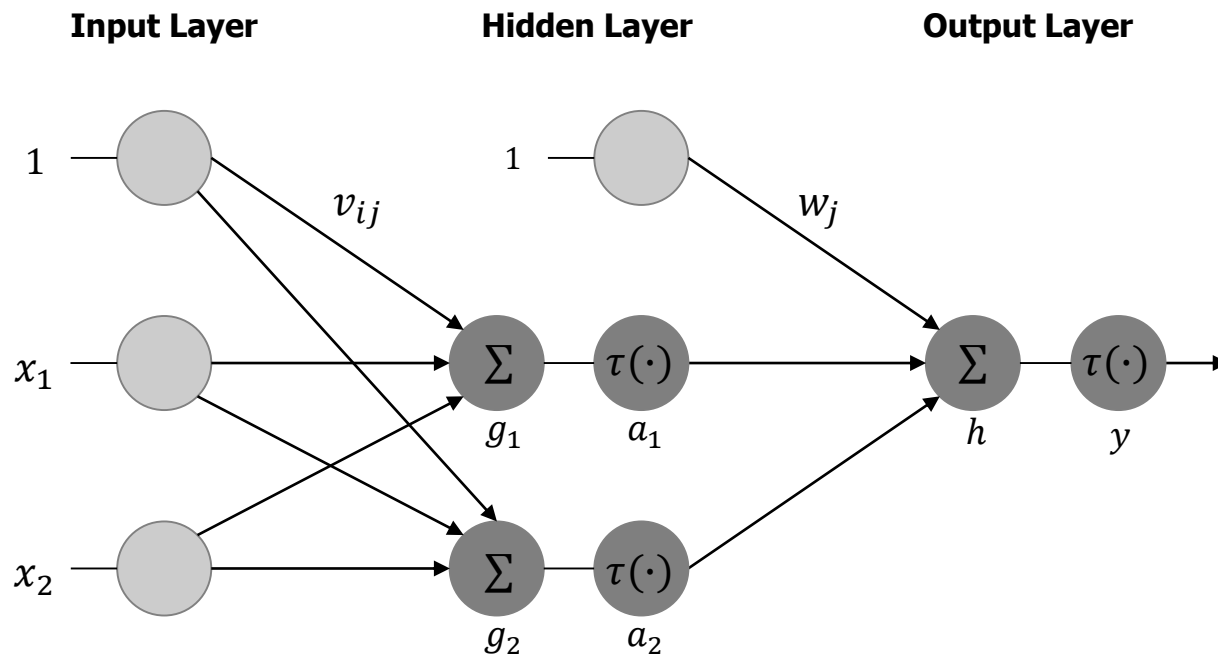
    print('it : %d  loss val : %.5f ' % (it, loss))

    vis_rt = (random_tensor + 1) / 2 * 255.0
    vis_rt = vis_rt.numpy()
    cv2.imshow('rt', vis_rt.astype(np.uint8))
    cv2.waitKey(1)
```

Neural Network

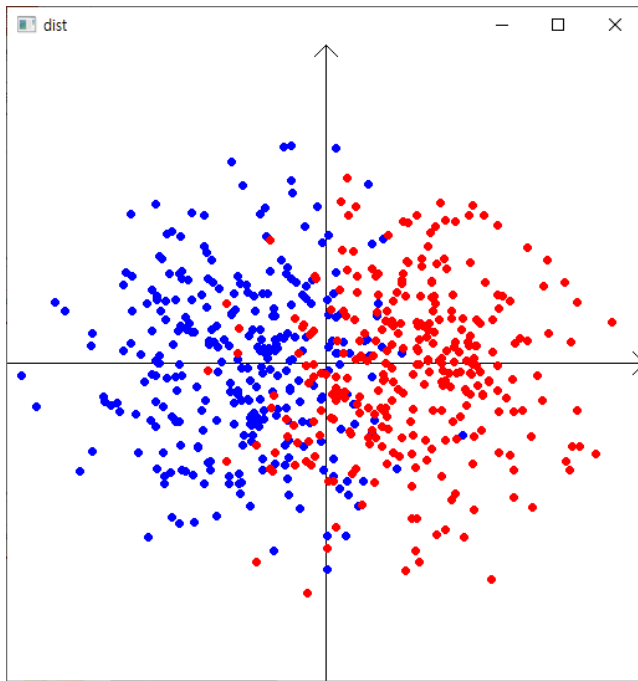
▪ Example of neural network

- Input, hidden, and out layers
- x_1 and x_2 : input values
- y : target value
- \mathbf{v}, \mathbf{w} : training values



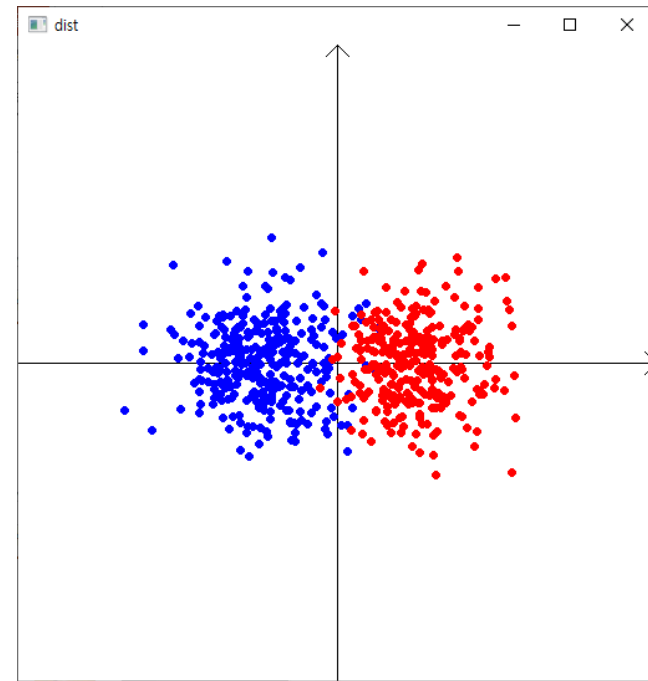
Neural Network

- Example of neural network (Single layer)
 - Step1. Build dataset and visualization
 - Two different Gaussian distributions



mean / std : (-1.0, 0.0) / (1.0, 1.0)

mean / std : (1.0, 0.0) / (1.0, 1.0)



mean / std : (-1.0, 0.0) / (0.5, 0.5)

mean / std : (1.0, 0.0) / (0.5, 0.5)

Neural Network

Example of neural net

- Step1. Building data
- Two different Gaussian

```
import numpy as np
import cv2

def gauss_2d(mu1, mu2, sigma=0.5, num=300):
    x = np.random.normal(mu1, sigma, num)
    y = np.random.normal(mu2, sigma, num)

    return (x, y)

def visualization(paper, dist, color):
    x = dist[0]
    y = dist[1]

    for it in range(len(x)):
        x_p = int(64 * x[it] + 256)
        y_p = int(-64 * y[it] + 256)
        cv2.circle(paper, (x_p, y_p), radius=2, color=color, thickness=2)

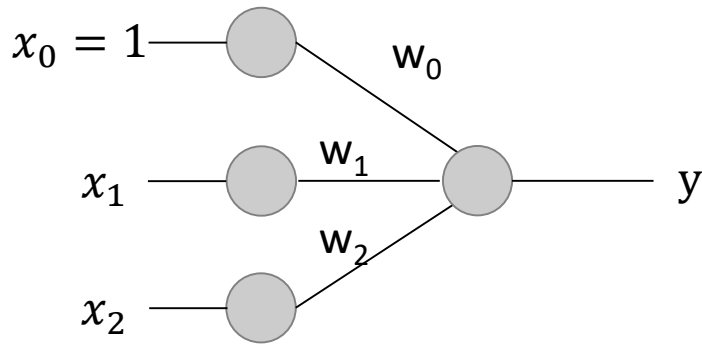
dist1 = gauss_2d(-1.0, 0.0)
dist2 = gauss_2d(1.0, 0.0)

paper = np.ones((512, 512, 3), dtype=np.uint8) * 255
paper = cv2.arrowedLine(paper, (0, 256), (512, 256), (0, 0, 0), tipLength=0.025)
paper = cv2.arrowedLine(paper, (256, 512), (256, 0), (0, 0, 0), tipLength=0.025)

visualization(paper, dist1, color=(255, 0, 0))
visualization(paper, dist2, color=(0, 0, 255))
cv2.imshow('dist', paper)
cv2.waitKey(-1)
```

Neural Network

- Example of neural network (Single layer)
 - Step2. Design network



```
# ---- network
class NeuralNet(torch.nn.Module):
    def __init__(self, input_size):
        super(NeuralNet, self).__init__()
        self.inpput_size = input_size
        self.fc1 = torch.nn.Linear(self.input_size, 1)
        self.act1 = torch.nn.Sigmoid()

    def forward(self, input_tensor):
        fc1 = self.fc1(input_tensor)
        output = self.act1(fc1)
        return output
```


Neural Network

▪ Example of neural network (Single layer)

- Step3. Training and test set

```
def train_test_set(mu_x1, mu_y1, mu_x2, mu_y2, sigma=0.5):  
    # ---- train / test  
    train = np.zeros((400, 2), dtype=np.float32)  
    train_gt = np.zeros((400, 1), dtype=np.float32)  
  
    test = np.zeros((200, 2), dtype=np.float32)  
    test_gt = np.zeros((200, 1), dtype=np.float32)  
  
    # ---- training set  
    for it in range(400):  
        if it < 200:  
            train[it, 0] = np.random.normal(mu_x1, sigma)  
            train[it, 1] = np.random.normal(mu_y1, sigma)  
            train_gt[it, 0] = 0  
        else:  
            train[it, 0] = np.random.normal(mu_x2, sigma)  
            train[it, 1] = np.random.normal(mu_y2, sigma)  
            train_gt[it, 0] = 1  
  
    # ---- test set  
    for it in range(200):  
        if it < 100:  
            test[it, 0] = np.random.normal(mu_x1, sigma)  
            test[it, 1] = np.random.normal(mu_y1, sigma)  
            test_gt[it, 0] = 0  
        else:  
            test[it, 0] = np.random.normal(mu_x2, sigma)  
            test[it, 1] = np.random.normal(mu_y2, sigma)  
            test_gt[it, 0] = 1  
  
    return train, train_gt, test, test_gt
```

Neural Network

- Example of neural network (Single layer)
 - Step4. Training and test network

```
from Lecture4_function import *
import torch

# ---- network
class NeuralNet(torch.nn.Module):
    def __init__(self, input_size):
        super(NeuralNet, self).__init__()
        self.input_size = input_size
        self.fc1 = torch.nn.Linear(self.input_size, 1)
        self.act1 = torch.nn.Sigmoid()

    def forward(self, input_tensor):
        fc1 = self.fc1(input_tensor)
        output = self.act1(fc1)
        return output

dist1 = gauss_2d(-1.0, 0.0)
dist2 = gauss_2d(1.0, 0.0)

paper = np.ones((512, 512, 3), dtype=np.uint8) * 255
paper = cv2.arrowedLine(paper, (0, 256), (512, 256), (0, 0, 0), tipLength=0.025)
paper = cv2.arrowedLine(paper, (256, 512), (256, 0), (0, 0, 0), tipLength=0.025)

visualization(paper, dist1, color=(255, 0, 0))
visualization(paper, dist2, color=(0, 0, 255))
cv2.imshow('dist', paper)
cv2.waitKey(1)
```

Neural Network

- Example of neural network
 - Step4. Training and test

```
# ----- network
model = NeuralNet(2)
learning_rate = 0.01
loss = torch.nn.BCELoss()
num_iter = 10000
optimizer = torch.optim.SGD(model.parameters(), lr=learning_rate)

train_set, train_gt, test_set, test_gt = train_test_set(-1.0, 0.0, 1.0, 0.0)

for it in range(num_iter):
    model.train()
    optimizer.zero_grad()
    train_output = model(torch.from_numpy(train_set))

    train_loss = loss(train_output, torch.from_numpy(train_gt))
    train_loss.backward()
    optimizer.step()

    model.eval()
    test_loss = loss(model(torch.from_numpy(test_set)), torch.from_numpy(test_gt))

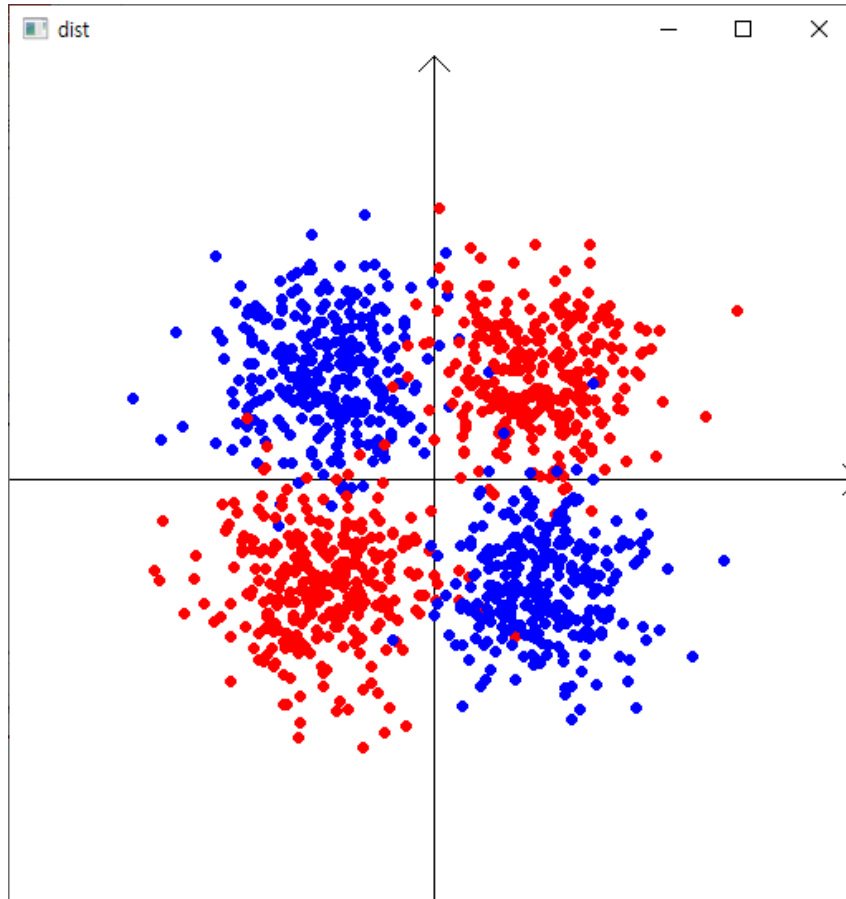
    if it % 100 == 0:
        print("train : %f  test : %f " % (train_loss.item(), test_loss.item()))

# ----- user input test
user_input = np.zeros((1, 2), dtype=np.float32)
user_input[0, 0] = 0.0
user_input[0, 1] = 0.0

user_output = model(torch.from_numpy(user_input))
print("user : %f " % user_output.item())
```

Neural Network

- Example of neural network (Multiple layers)
 - XOR Problem



mean1 / mean2: (-1.0, 1.0) / (1.0, -1.0)

mean1 / mean2: (1.0, 1.0) / (-1.0, -1.0)

Neural Network

- Example of neural network (Multiple layers)
 - Training and test set

```
def train_test_set(mu_x1, mu_y1, mu_x2, mu_y2, mu_x3, mu_y3, mu_x4, mu_y4, sigma=0.5):
    # ---- train / test
    train = np.zeros((400, 2), dtype=np.float32)
    train_gt = np.zeros((400, 1), dtype=np.float32)

    test = np.zeros((200, 2), dtype=np.float32)
    test_gt = np.zeros((200, 1), dtype=np.float32)

    # ---- training set
    for it in range(400):
        if it < 100:
            train[it, 0] = np.random.normal(mu_x1, sigma)
            train[it, 1] = np.random.normal(mu_y1, sigma)
            train_gt[it, 0] = 0
        elif it >= 100 and it < 200:
            train[it, 0] = np.random.normal(mu_x2, sigma)
            train[it, 1] = np.random.normal(mu_y2, sigma)
            train_gt[it, 0] = 1
        elif it >= 200 and it < 300:
            train[it, 0] = np.random.normal(mu_x3, sigma)
            train[it, 1] = np.random.normal(mu_y3, sigma)
            train_gt[it, 0] = 1
        else:
            train[it, 0] = np.random.normal(mu_x4, sigma)
            train[it, 1] = np.random.normal(mu_y4, sigma)
            train_gt[it, 0] = 0

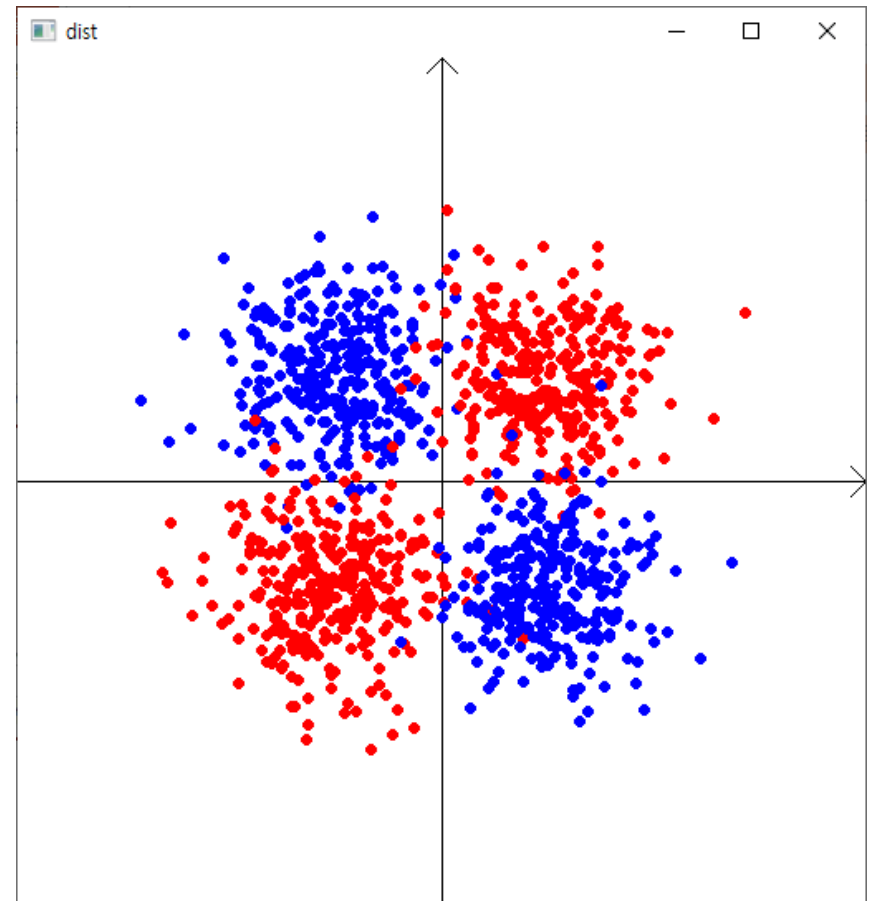
    # ---- test set
    for it in range(200):
        if it < 50:
            test[it, 0] = np.random.normal(mu_x1, sigma)
            test[it, 1] = np.random.normal(mu_y1, sigma)
            test_gt[it, 0] = 0
        elif it >= 50 and it < 100:
            test[it, 0] = np.random.normal(mu_x2, sigma)
            test[it, 1] = np.random.normal(mu_y2, sigma)
            test_gt[it, 0] = 1
        elif it >= 100 and it < 150:
            test[it, 0] = np.random.normal(mu_x3, sigma)
            test[it, 1] = np.random.normal(mu_y3, sigma)
            test_gt[it, 0] = 1
        else:
            test[it, 0] = np.random.normal(mu_x4, sigma)
            test[it, 1] = np.random.normal(mu_y4, sigma)
            test_gt[it, 0] = 0

    return train, train_gt, test, test_gt
```

Neural Network

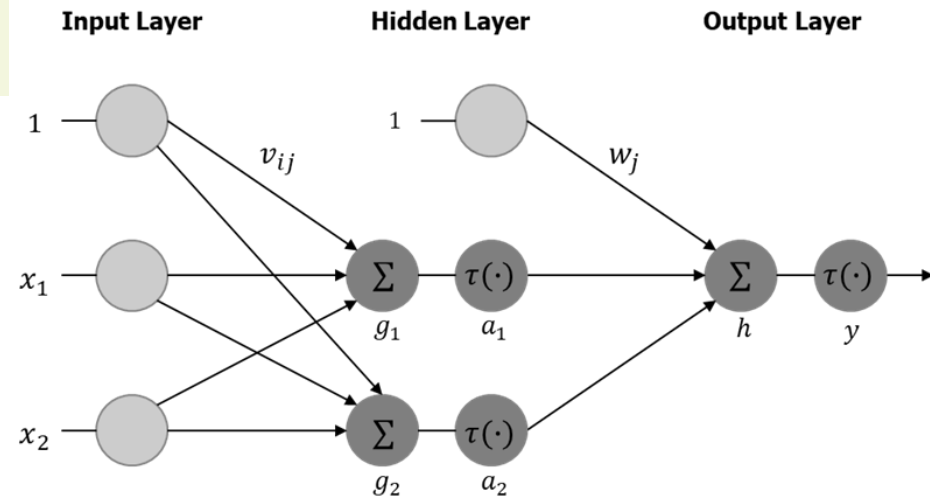
- Example of neural network (Multiple layers)
 - Training Network – **Fail**

```
Lecture_04 ×  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
train : 0.691956 test : 0.692072  
user : 0.507929  
  
Process finished with exit code 0
```



Neural Network

- Example of neural network (Multiple layers)
 - Redesign network



```
# ---- network
class NeuralNet(torch.nn.Module):
    def __init__(self, input_size, hidden_size):
        super(NeuralNet, self).__init__()
        self.input_size = input_size
        self.hidden_size = hidden_size
        self.fc1 = torch.nn.Linear(self.input_size, self.hidden_size)
        self.act1 = torch.nn.ReLU()
        self.fc2 = torch.nn.Linear(self.hidden_size, 1)
        self.act2 = torch.nn.Sigmoid()

    def forward(self, input_tensor):
        fc1 = self.fc1(input_tensor)
        act1 = self.act1(fc1)
        fc2 = self.fc2(act1)
        output = self.act2(fc2)
        return output
```

Neural Network

- Example of neural network (Multiple layers)
 - Redesign network
 - Hidden layer : 2 / 5 / 10

```
Run: Lecture_04 ×
train : 0.271659 test : 0.346585
train : 0.271642 test : 0.346621
train : 0.271625 test : 0.346628
train : 0.271609 test : 0.346629
train : 0.271594 test : 0.346644
train : 0.271580 test : 0.346651
train : 0.271566 test : 0.346657
train : 0.271552 test : 0.346660
train : 0.271538 test : 0.346661
train : 0.271525 test : 0.346661
train : 0.271512 test : 0.346658
train : 0.271499 test : 0.346655
train : 0.271487 test : 0.346650
train : 0.271475 test : 0.346646
train : 0.271463 test : 0.346666
user : 0.005645
```

```
Lecture_04 ×
train : 0.095448 test : 0.132818
train : 0.095423 test : 0.132818
train : 0.095400 test : 0.132852
train : 0.095378 test : 0.132873
train : 0.095356 test : 0.132884
train : 0.095334 test : 0.132891
train : 0.095313 test : 0.132894
train : 0.095292 test : 0.132897
train : 0.095272 test : 0.132898
train : 0.095251 test : 0.132899
train : 0.095232 test : 0.132898
train : 0.095212 test : 0.132896
train : 0.095193 test : 0.132894
train : 0.095175 test : 0.132890
train : 0.095156 test : 0.132886
user : 0.000228
```

```
Lecture_04 ×
train : 0.089703 test : 0.100805
train : 0.089699 test : 0.100828
train : 0.089695 test : 0.100851
train : 0.089691 test : 0.100873
train : 0.089687 test : 0.100895
train : 0.089683 test : 0.100917
train : 0.089680 test : 0.100937
train : 0.089676 test : 0.100958
train : 0.089672 test : 0.100978
train : 0.089669 test : 0.100992
train : 0.089665 test : 0.101009
train : 0.089662 test : 0.101026
train : 0.089659 test : 0.101042
train : 0.089656 test : 0.101058
train : 0.089653 test : 0.101074
user : 0.000450
```


Neural Network

- Practice : Add one more layer

