```
import numpy as np

## AND 연산
inputs = [(0,0),(0,1),(1,0),(1,1)] # 튜플로 x 설정
w1, w2, b = 0.5,0.5,-0.7 # 가중치와 y절편(bias)
```

```
inputs = [(0,0),(0,1),(1,0),(1,1)] # 튜플로 x 설정
w1, w2, b = 0.5,0.5,-0.7 # 가중치와 y절편(bias)
for x1, x2 in inputs :
  y = w1*x1 + w2*x2 + b
  if y <= 0 :
    y = 0
  else :
    y = 1

print('({x1},{x2}) => {y}'.format(x1=x1, x2=x2, y=y))
```

```
\begin{array}{ccc} (0,0) \Rightarrow 0 \\ (0,1) \Rightarrow 0 \\ (1,0) \Rightarrow 0 \\ (1,1) \Rightarrow 1 \end{array}
```

```
## NAND 연산 : 가중치 -0.5, bias = -0.7
## OR 연산 : 가중치 0.5, bias = -0.2
```

```
def AND(x1, x2):
 x = np.array([x1,x2])
 w = np.array([0.5, 0.5])
 b = -0.7
 y = np.sum(x*w) + b
 if y \le 0:
   return 0
 else :
    return 1
def NAND(x1, x2):
  x = np.array([x1,x2])
 w = np.array([-0.5, -0.5])
 b = 0.7
 y = np.sum(x*w)+b
 if y \le 0:
  return 0
  else:
   return 1
def OR(x1, x2):
 x = np.array([x1,x2])
 w = np.array([0.5, 0.5])
 b = -0.2
 y = np.sum(x*w)+b
 if y \le 0:
    return 0
  else:
    return 1
```

```
## multi perceptron ##
def XOR(x1,x2):
  s1 = NAND(x1,x2)
  s2 = OR(x1, x2)
  y = AND(s1,s2)
  return y
inputs =([0,0],[1,0],[0,1],[1,1])
print('AND :')
for x1, x2 in inputs:
  y1 = AND(x1,x2)
  print('(\{x1\},\{x2\}) \rightarrow \{y\}'.format(x1=x1, x2=x2, y=y1))
print('NAND :')
for x1, x2 in inputs :
  y = NAND(x1,x2)
  print('(\{x1\},\{x2\}) \rightarrow \{y\}'.format(x1=x1, x2=x2, y=y))
print('OR :')
for x1, x2 in inputs :
  y = OR(x1, x2)
  print('(\{x1\},\{x2\}) -> \{y\}'.format(x1=x1,x2=x2,y=y))
print('XOR :')
for x1, x2 in inputs:
  y = XOR(x1,x2)
  print('({x1},{x2}) \rightarrow {y}'.format(x1=x1, x2=x2, y=y))
 AND :
       (0,0) \rightarrow 0
       (1,0) \rightarrow 0
       (0,1) \rightarrow 0
       (1,1) \rightarrow 1
       NAND :
       (0,0) \rightarrow 1
       (1,0) \rightarrow 1
       (0,1) \rightarrow 1
       (1,1) \rightarrow 0
       OR :
       (0,0) \rightarrow 0
       (1,0) \rightarrow 1
       (0,1) \rightarrow 1
       (1,1) \to 1
       XOR :
       (0,0) \rightarrow 0
       (1,0) \rightarrow 1
       (0,1) \rightarrow 1
       (1,1) \rightarrow 0
```

CS231N

https://tykimos.github.io/lecture/

```
# 역전파
# optimizer : 가중치를 찾는 과정
# 새 가중치 = 기존 가중치 - n(learning rate) * 편미분 E/w(gradient)
# gradient 어떤 방향으로 학습?
# learning rate 얼마나 학습시킬지
```

0