

✧ XOR Problem

```
import numpy as np
import matplotlib.pyplot as plt

from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
from scipy import optimize
```

✧ Bitwise Operator Dataset

```
# XOR
X = np.array([[0,0], [0,1], [1,0], [1,1]])
y = np.array([0,1,1,0]).flatten()

# OR
# X = np.array([[0,0], [0,1], [1,0], [1,1]])
# y = np.array([0,1,1,1]).flatten()

# AND
# X = np.array([[0,0], [0,1], [1,0], [1,1]])
# y = np.array([0,0,0,1]).flatten()
```

✧ Perceptron

```
def perceptron(X, W, b):
    ## IMPLEMENT HERE
    y = X@W + b
    y = sigmoid(y)
    return y

def sigmoid(z):
    ## IMPLEMENT HERE
    z = 1 / (1 + np.exp(-z))
    return z
```

```
np.log(0+1e-10) # -inf 음의 무한대 값을 없애기 위해서 -최소값을 넣어서 해결
```

```
→ -23.025850929940457
```

✧ Single Perceptron Learning

```

def bce_loss(weights, args):
    ## IMPLEMENT HERE
    X = args[0]
    y = args[1]

    W, b = weights[:-1], weights[-1]
    y_hat = perceptron(X, W, b)
    bce = -y * np.log(y_hat) - (1.0 - y) * np.log(1.0 - y_hat + 1e-10)
    los = bce.mean()
    return los

result = optimize.minimize(fun = bce_loss, x0 = [0, 0, 0], args=[X, y])

W_opt, b_opt = result.x[:-1], result.x[-1]
y_hat = perceptron(X, W_opt, b_opt)
y_hat_cls = (y_hat > 0.5).astype('int8')
print(y_hat_cls)

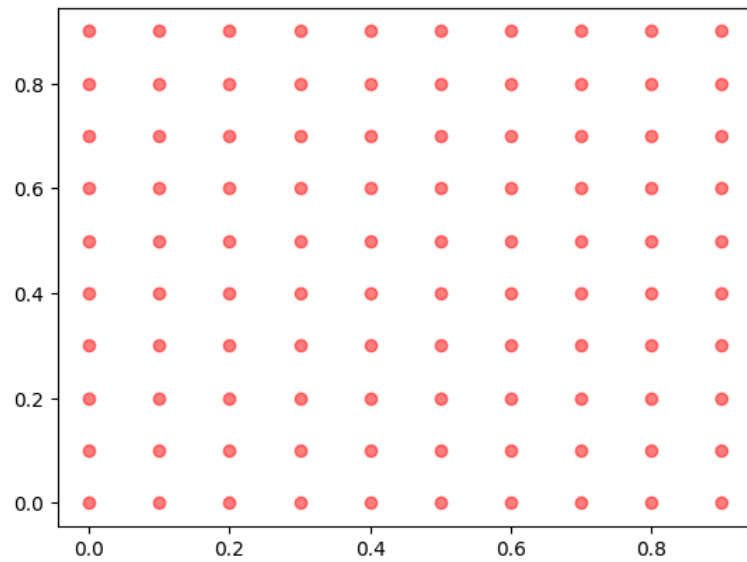
accuracy = accuracy_score(y, y_hat_cls)
print(f"Accuracy: {accuracy}")

↩ [0 0 0 0]
  Accuracy: 0.5

## visualization
color = ['red', 'blue']
for x0 in np.arange(0,1,0.1):
    for x1 in np.arange(0,1,0.1):
        x = np.array([x0, x1])
        y_hat = perceptron(x, W_opt, b_opt)
        y_hat_cls = (y_hat > 0.5).astype('int8')
        plt.scatter(x0, x1, c=color[y_hat_cls], alpha=0.5)

plt.show()

```



✓ Multiple Perceptrons Learning

```
w11 = np.array([5, 5])
w12 = np.array([-7, -7])
w2 = np.array([-15, -15])
b1 = -8
b2 = 3
b3 = 6

def predict(x):
    ## IMPLEMENT HERE
    y1 = perceptron(x, w11, b1)
    y2 = perceptron(x, w12, b2)
    x2 = np.array([y1, y2])
    y_hat = perceptron(x2, w2, b3)
    return y_hat, x2
```

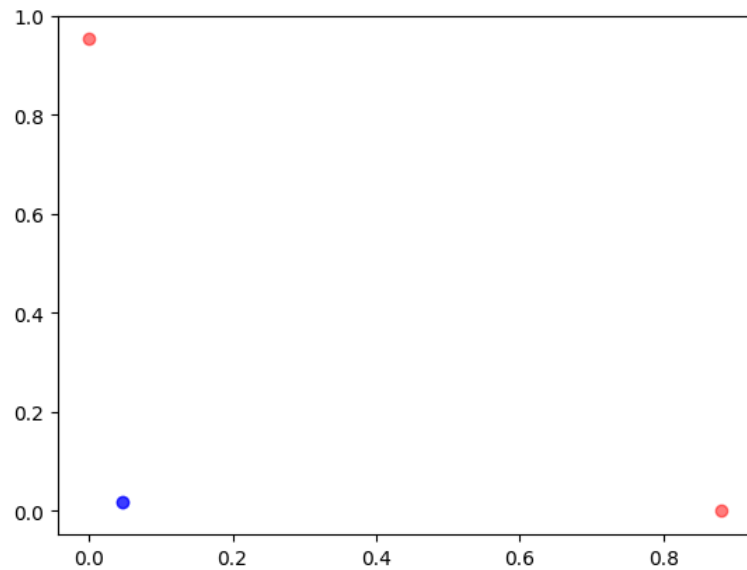
✓ Analysis of multiple perceptron

```
# multiple perceptron

for x in X:
    y_hat, layer_x2 = predict(x)
    y_hat_cls = (y_hat > 0.5).astype('int8')
    print(f'{x} => {y_hat_cls}')
```

```
plt.scatter(layer_x2[0], layer_x2[1], c=color[y_hat_cls], alpha=0.5)
```

```
⇒ [0 0] => 0  
[0 1] => 1  
[1 0] => 1  
[1 1] => 0
```



```
color = ['red', 'blue']  
for x0 in np.arange(0,1,0.1):  
    for x1 in np.arange(0,1,0.1):  
        x = np.array([x0, x1])  
        y_hat, layer_x2 = predict(x)  
        y_hat_cls = (y_hat > 0.5).astype('int8')  
        plt.scatter(x0, x1, c=color[y_hat_cls], alpha=0.5)  
  
plt.show()
```

