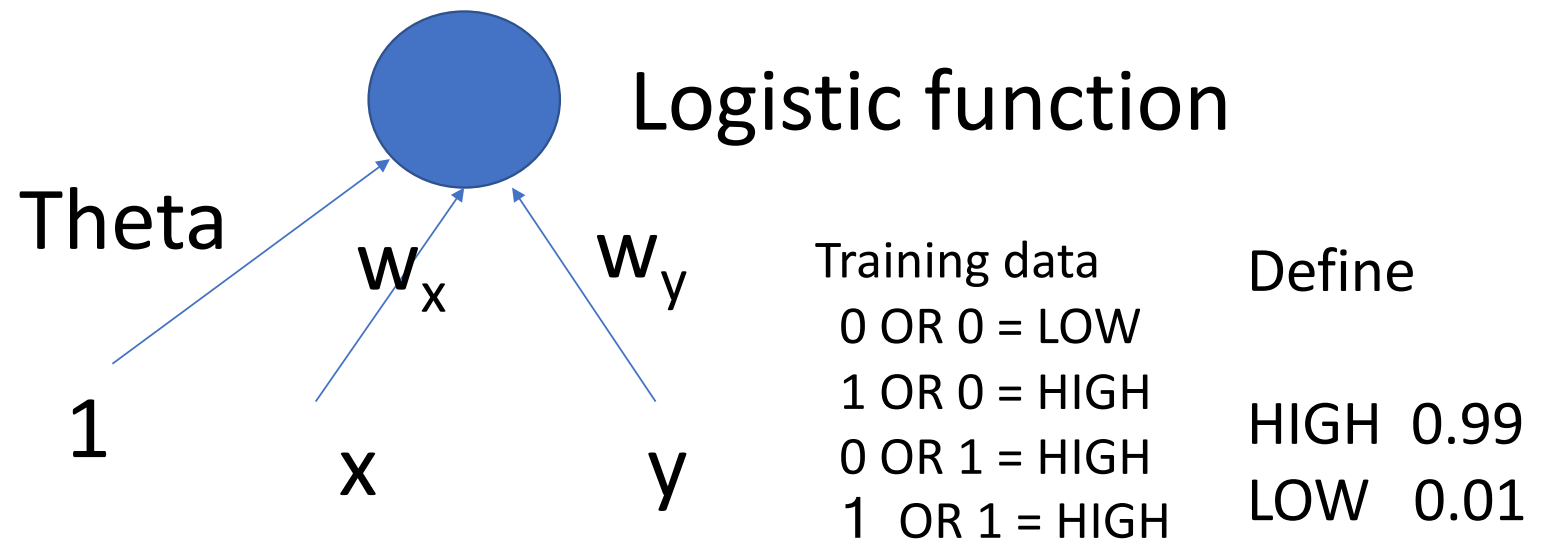


# Perceptron OR gate

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# Single perceptron to implement OR



# Perceptron training

- Train a perceptron to do OR operation
- Logic OR gate with input polluted with Gaussian noise
- there are four type inputs
  - #  $0 \text{ OR } 0 = \text{LOW}$
  - #  $1 \text{ OR } 0 = \text{HIGH}$
  - #  $0 \text{ OR } 1 = \text{HIGH}$
  - #  $1 \text{ OR } 1 = \text{HIGH}$

# Delta rule

- Generate a sample for input (0,0) and update
- Generate a sample for input (1,0) and update
- Generate a sample for input (0,1) and update
- Generate a sample for input (1,1) and update

Generate a sample for input (0,0)

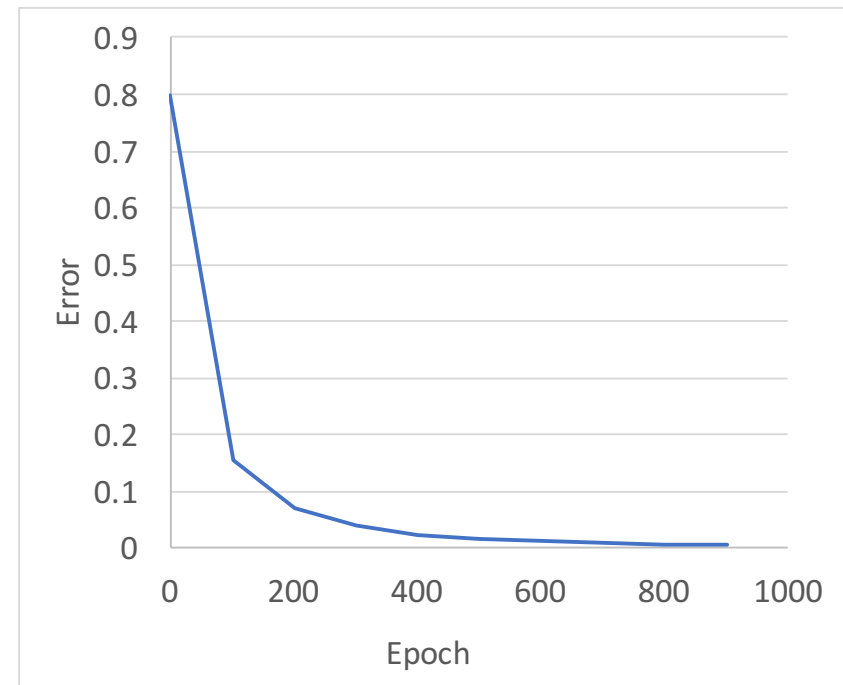
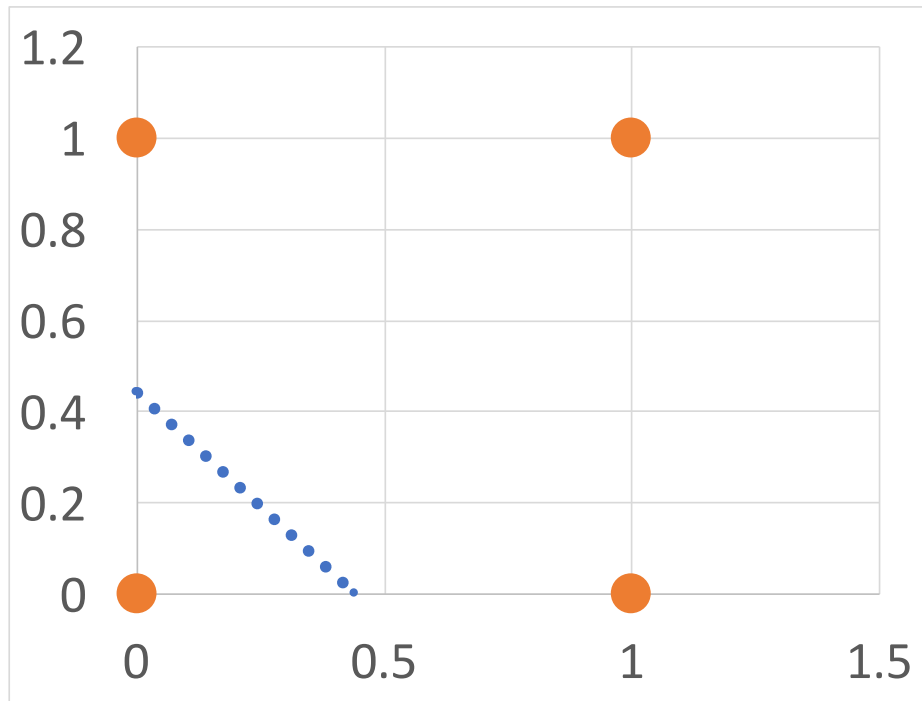
```
x00=random.gauss(0,0.1)
y00=random.gauss(0,0.1)
output_cal=logistic(w_x,w_y,theta, x00,y00)
#delta rule
w_x=w_x-eta*(output_cal-L0W)*x00
w_y=w_y-eta*(output_cal-L0W)*y00
theta= theta-eta*(output_cal-L0W)*1
```

Delta rule:

Update = -Learning rate(Actual Output-Desired output)

# Final weight coefficient

wx 6.09 wy 6.09 theta -2.693



Test the neuron for input

```
x 0 y 0 output 0.06336497914983781|  
x 1 y 0 output 0.9677346756607005  
x 0 y 1 output 0.9677031963574627  
x 1 y 1 output 0.9999247262916799
```

```

from math import *
print('exp(0)', round(exp(1),4)) #verify exponential function is working

eta=0.1 #define learning rate
# initialization of weight coefficient
w_x=random.uniform(-0.1,0.1)
w_y=random.uniform(0.1,0.1)
theta=random.uniform(-0.1,0.1)

#define HIGH and LOW
HIGH= 0.99
LOW = 0.01

def logistic(w_x,w_y, theta, x,y):
    return 1/(1+exp(-w_x*x-w_y*y-theta))

#feed in training sample type 1
#calcuated expected output

for i in range(1000):
    x00=random.gauss(0,0.1)
    y00=random.gauss(0,0.1)
    output_cal=logistic(w_x,w_y,theta, x00,y00)
    #delta rule
    w_x=w_x-eta*(output_cal-LOW)*x00
    w_y=w_x-eta*(output_cal-LOW)*y00
    theta= theta-eta*(output_cal-LOW)*1

    x10=random.gauss(1,0.1)
    y10=random.gauss(0,0.1)

    output_cal=logistic(w_x,w_y,theta, x10,y10)
    #delta rule
    w_x=w_x-eta*(output_cal-HIGH)*x10    #change here for other logic gate
    w_y=w_x-eta*(output_cal-HIGH)*y10
    theta= theta-eta*(output_cal-HIGH)*1

    x01=random.gauss(0,0.1)
    y01=random.gauss(1,0.1)

    output_cal=logistic(w_x,w_y,theta, x01,y01)
    #delta rule
    w_x=w_x-eta*(output_cal-HIGH)*x01 #change here for other logic gate
    w_y=w_x-eta*(output_cal-HIGH)*y01
    theta= theta-eta*(output_cal-HIGH)*1

```



```

x11=random.gauss(1,0.1)
y11=random.gauss(1,0.1)

output_cal=logistic(w_x,w_y,theta, x11,y11)
#delta rule
w_x=w_x-eta*(output_cal-HIGH)*x11
w_y=w_y-eta*(output_cal-HIGH)*y11
theta= theta-eta*(output_cal-HIGH)*1
#sampling the error every 100 sample
if i % 100==0:    #index divided by 10 = 0
    # error calculation
    error=0
    x_test=0
    y_test=0
    error= error+(logistic(w_x, w_y,theta, x_test, y_test)-LOW)**2  #for OR expected outcome LOW

    x_test=1
    y_test=0
    error= error+(logistic(w_x, w_y,theta, x_test, y_test)-HIGH)**2  #for OR expected outcome HIGH

    x_test=0
    y_test=1
    error= error+(logistic(w_x, w_y,theta, x_test, y_test)-HIGH)**2  #for OR expected outcome HIGH

    x_test=1
    y_test=1
    error= error+(logistic(w_x, w_y,theta, x_test, y_test)-HIGH)**2  #for OR expected outcome HIGH
print('iteration',i, 'error', error)

```

```
print('wx', w_x, 'wy', w_y, 'theta', theta)
# calculate expect output
x_test=0
y_test=0
test_output= logistic(w_x, w_y,theta, x_test, y_test)
print('x', x_test, 'y', y_test, 'output', test_output)

x_test=1
y_test=0
test_output= logistic(w_x, w_y,theta, x_test, y_test)
print('x', x_test, 'y', y_test, 'output', test_output)

x_test=0
y_test=1
test_output= logistic(w_x, w_y,theta, x_test, y_test)
print('x', x_test, 'y', y_test, 'output', test_output)
x_test=1
y_test=1
test_output= logistic(w_x, w_y,theta, x_test, y_test)
print('x', x_test, 'y', y_test, 'output', test_output)
```