

Chapter 3 Neural Network

3.3 Calculation of Multi-dimensional array

3.4 Implementation of 3-layer Neural Network

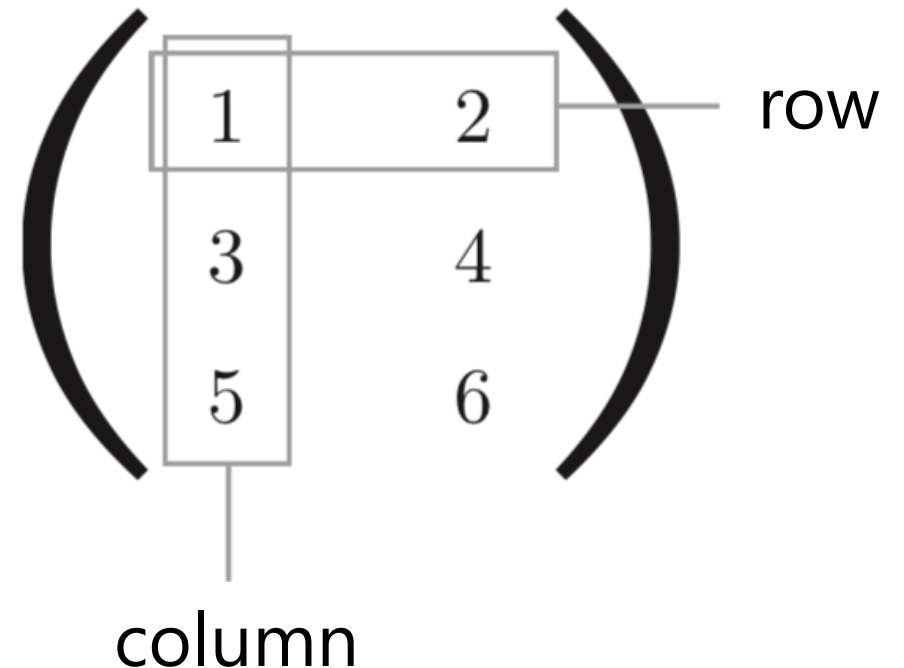
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Topics

- Multi-dimensional Array
- Product of matrix
- Implementation of Simple Neural Network
- Summary of implementation
- Summary

Multi-dimensional Array

- "Set of number"
- to place in a row, rectangular, 3-dimensional, ... , N-dimensional shape
- 2-dimensional array is called "matrix"



Create an Array using NumPy

※NumPy (Number + Python) ⇒ “ナンパイ” or ナムパイ

```
>>> import numpy as np
```

```
>>> A = np.array([1, 2, 3, 4]) ⇒ Create an 1-dimensional array
```

```
>>> print(A)
```

```
[1 2 3 4]
```

```
>>> np.ndim(A) ⇒ Acquire dimension of array
```

```
1
```

```
>>> A.shape ⇒ Acquire shape of array
```

```
(4,)
```

```
>>> A.shape[0] ⇒ Acquire number of elements
```

```
4
```

(1 2 3 4)

Create an Array using NumPy

```
>>> B = np.array([[1,2], [3,4], [5,6]]) ⇒ Create a 2-dimensional array(matrix)
>>> print(B)
[[1 2]
 [3 4]
 [5 6]]
>>> np.ndim(B) ⇒ Acquire dimension of array
2
>>> B.shape ⇒ Acquire shape of array
(3, 2)
```

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix}$$

Product of matrix

```
>>> A = np.array([[1,2], [3,4]]) ⇒ Create a 2-dimensional array(matrix)
>>> A.shape ⇒ Acquire shape
(2, 2)
>>> B = np.array([[5,6], [7,8]]) ⇒ Create a 2-dimensional array(matrix)
>>> B.shape ⇒ Acquire shape
(2, 2)
>>> np.dot(A, B) ⇒ Calculate product of matrix
array([[19, 22],
       [43, 50]])
```

np.dot()
⇒ can take at most 3 arguments

↓ revise in 5/21

np.dot(A,B,C) ⇒ not means "A · B · C"
substitute result of "A · B" for C

The diagram illustrates the dot product of two 2x2 matrices, A and B, resulting in a 2x2 matrix C. Matrix A is $\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$ and Matrix B is $\begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix}$. The resulting matrix C is $\begin{pmatrix} 19 & 22 \\ 43 & 50 \end{pmatrix}$. The calculation for the first row of C is shown: $1 \times 5 + 2 \times 7 = 19$ and $1 \times 6 + 2 \times 8 = 22$. The calculation for the second row of C is shown: $3 \times 5 + 4 \times 7 = 43$ and $3 \times 6 + 4 \times 8 = 50$. Arrows connect the elements of A and B to the corresponding elements in C, with the dot product formula for each element.

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} = \begin{pmatrix} 19 & 22 \\ 43 & 50 \end{pmatrix}$$

Labels A and B are placed below the first and second matrices respectively. The calculation for the first row of C is shown above the first row of C: $1 \times 5 + 2 \times 7$ and $1 \times 6 + 2 \times 8$. The calculation for the second row of C is shown below the second row of C: $3 \times 5 + 4 \times 7$ and $3 \times 6 + 4 \times 8$.

Error in product of matrix

```
>>> C.shape
```

```
(2, 2)
```

```
>>> A.shape
```

```
(2, 3)
```

```
>>> np.dot(A, C)
```

```
Traceback (most recent call last):
```

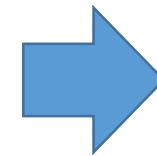
```
File "<stdin>", line 1, in <module>
```

```
ValueError: shapes (2,3) and (2,2) not aligned: 3 (dim 1) != 2 (dim 0)
```

A

C

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$$



Not calculable

Calculable example

A

B

=

C

shape: 3 × 2

2 × 4

3 × 4



Calculable

match the number of elements

Product of matrix in 1-layer Neural Network

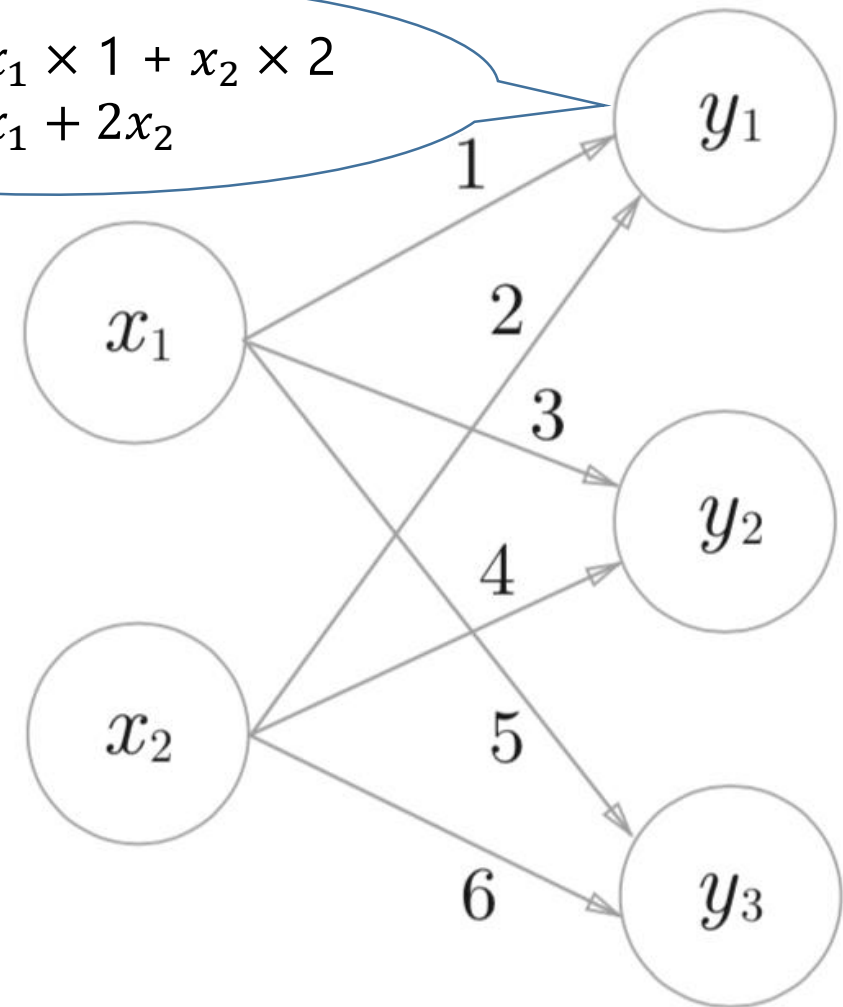
```
>>> X = np.array([1, 2])  
>>> X.shape  
(2,)  
>>> W = np.array([[1, 3, 5], [2, 4, 6]])  
>>> print(W)  
[[1 3 5]  
 [2 4 6]]  
>>> W.shape  
(2, 3)  
>>> Y = np.dot(X, W)  
>>> print(Y)  
[ 5 11 17]
```

$X = (x_1, x_2) \cdot \cdot \cdot$ Input

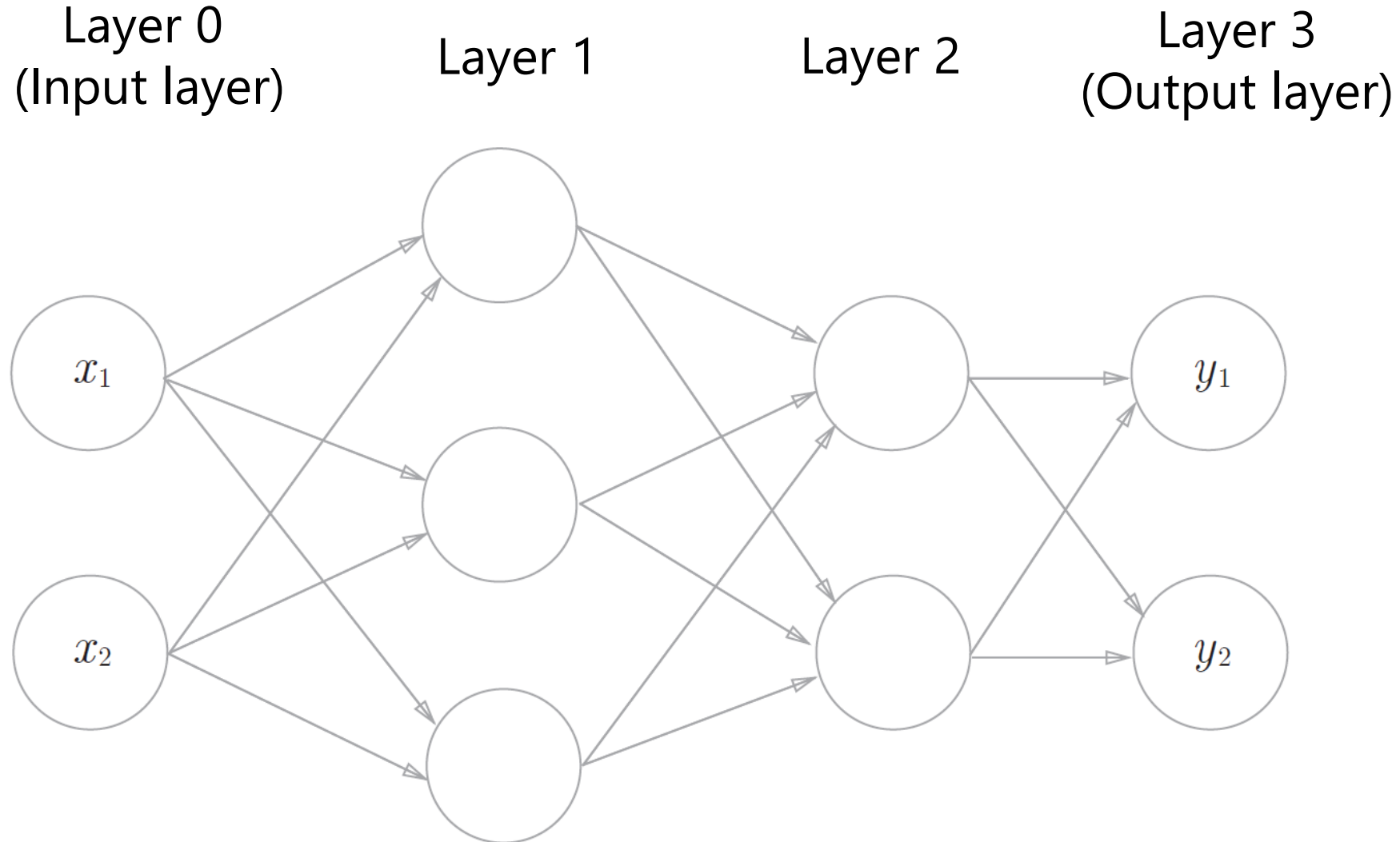
$W = \begin{pmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{pmatrix} \cdot \cdot \cdot$ Weight

$Y = (y_1, y_2, y_3) \cdot \cdot \cdot$ Output

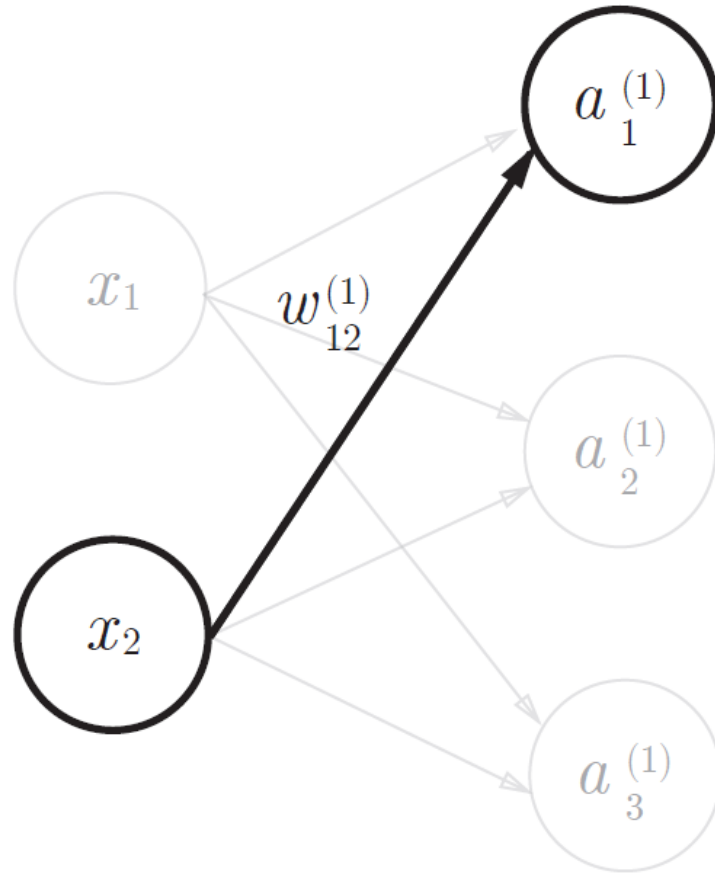
ex.) $y_1 = x_1 \times 1 + x_2 \times 2$
 $= x_1 + 2x_2$



Implementation of 3-layer Neural Network



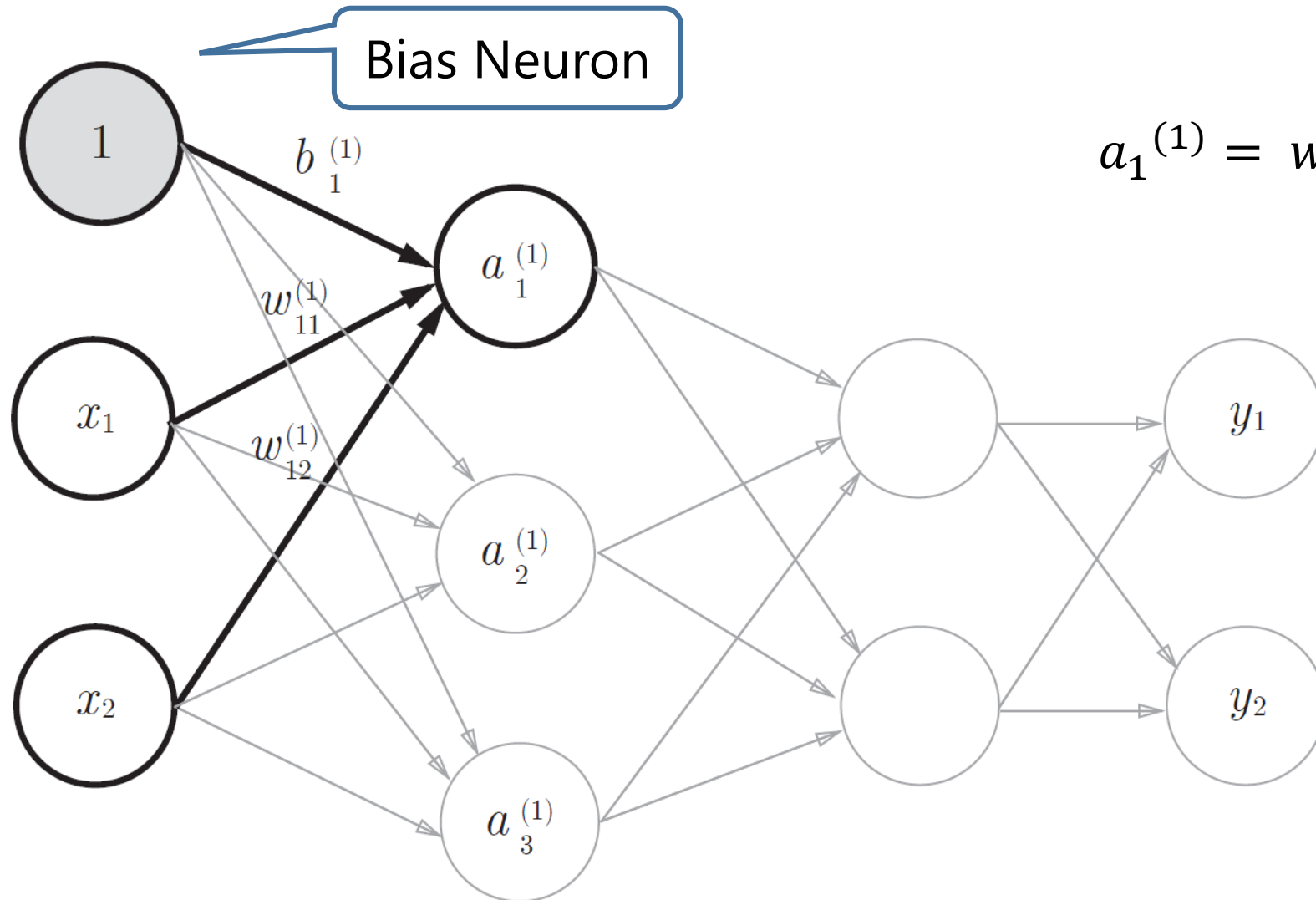
Definition of symbol



$w_{12}^{(1)}$ Weight in 1st layer
2nd neuron in **front** layer
1st neuron in **next** layer

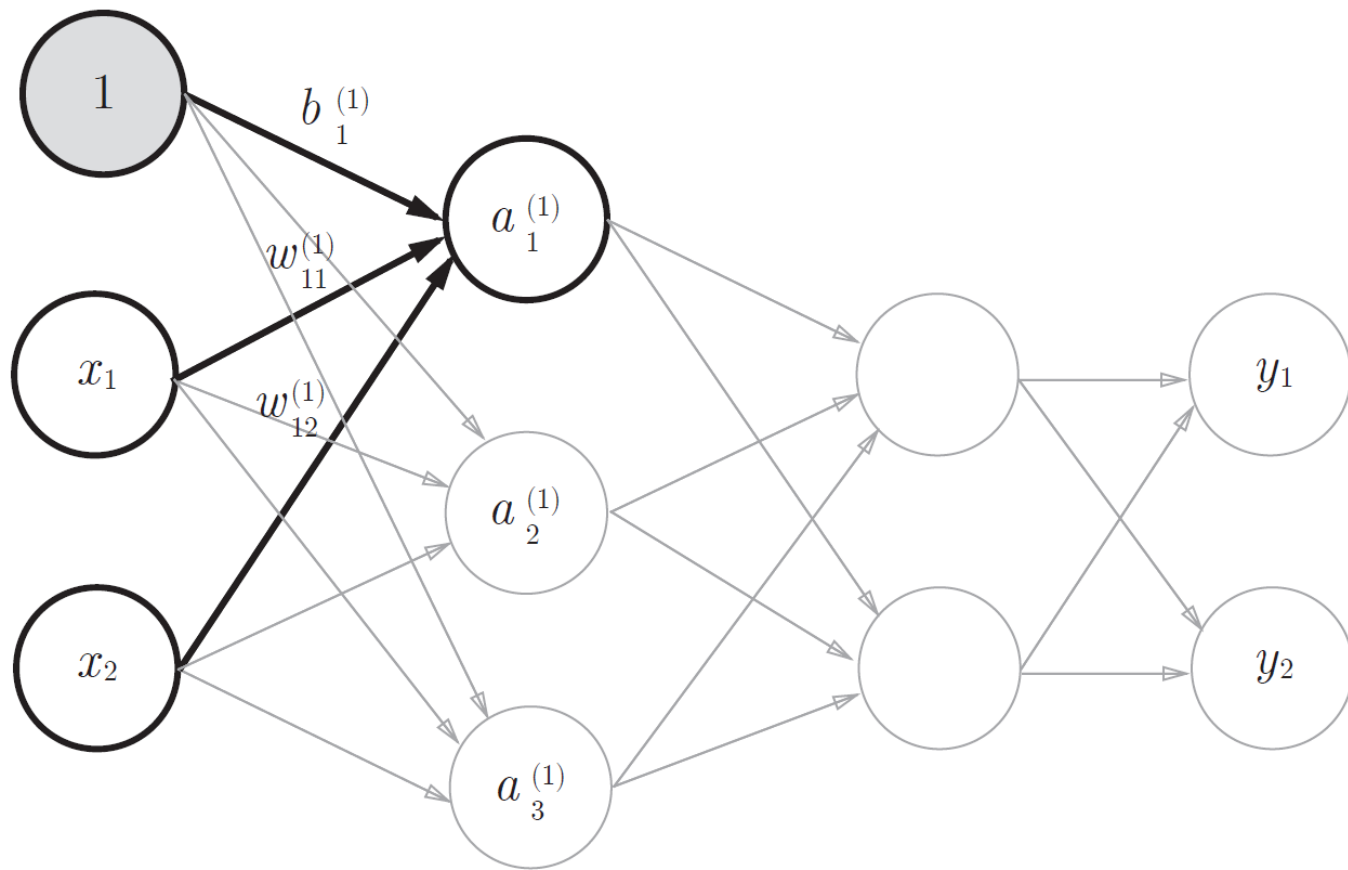
$a_1^{(1)}$ Neuron in 1st layer
1st neuron in this layer

Signal transmission to 1st neuron in 1st layer



$$a_1^{(1)} = w_{11}^{(1)}x_1 + w_{12}^{(1)}x_2 + b_1^{(1)}$$

Signal transmission to 1st layer



$$A^{(1)} = (a_1^{(1)} \ a_2^{(1)} \ a_3^{(1)})$$

$$X = (x_1 \ x_2)$$

$$B^{(1)} = (b_1^{(1)} \ b_2^{(1)} \ b_3^{(1)})$$

$$W^{(1)} = \begin{pmatrix} w_{11}^{(1)} & w_{21}^{(1)} & w_{31}^{(1)} \\ w_{12}^{(1)} & w_{22}^{(1)} & w_{32}^{(1)} \end{pmatrix}$$

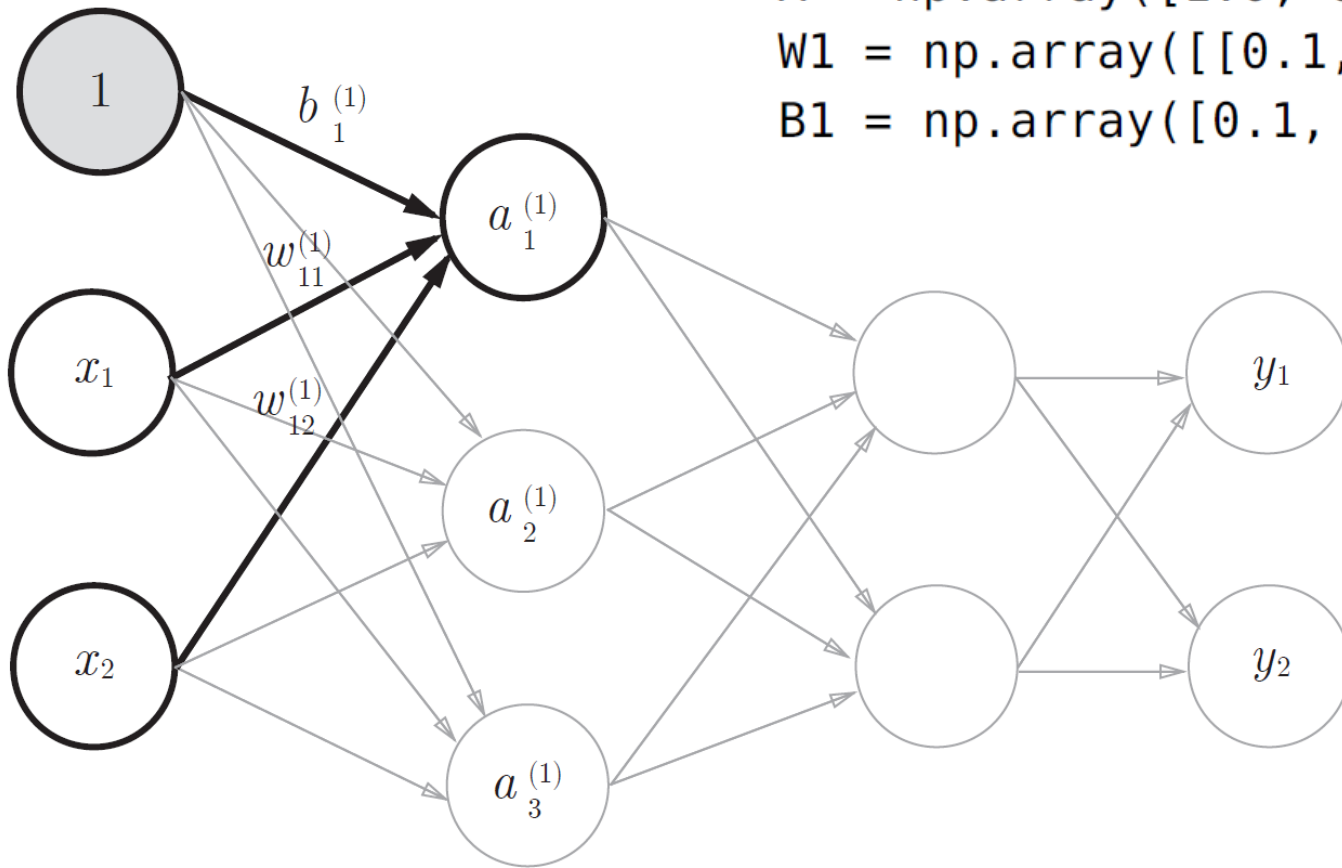
$$A^{(1)} = XW^{(1)} + B^{(1)}$$



Calculable at once

Signal transmission to 1st layer using NumPy

```
X = np.array([1.0, 0.5])  
W1 = np.array([[0.1, 0.3, 0.5], [0.2, 0.4, 0.6]])  
B1 = np.array([0.1, 0.2, 0.3])
```

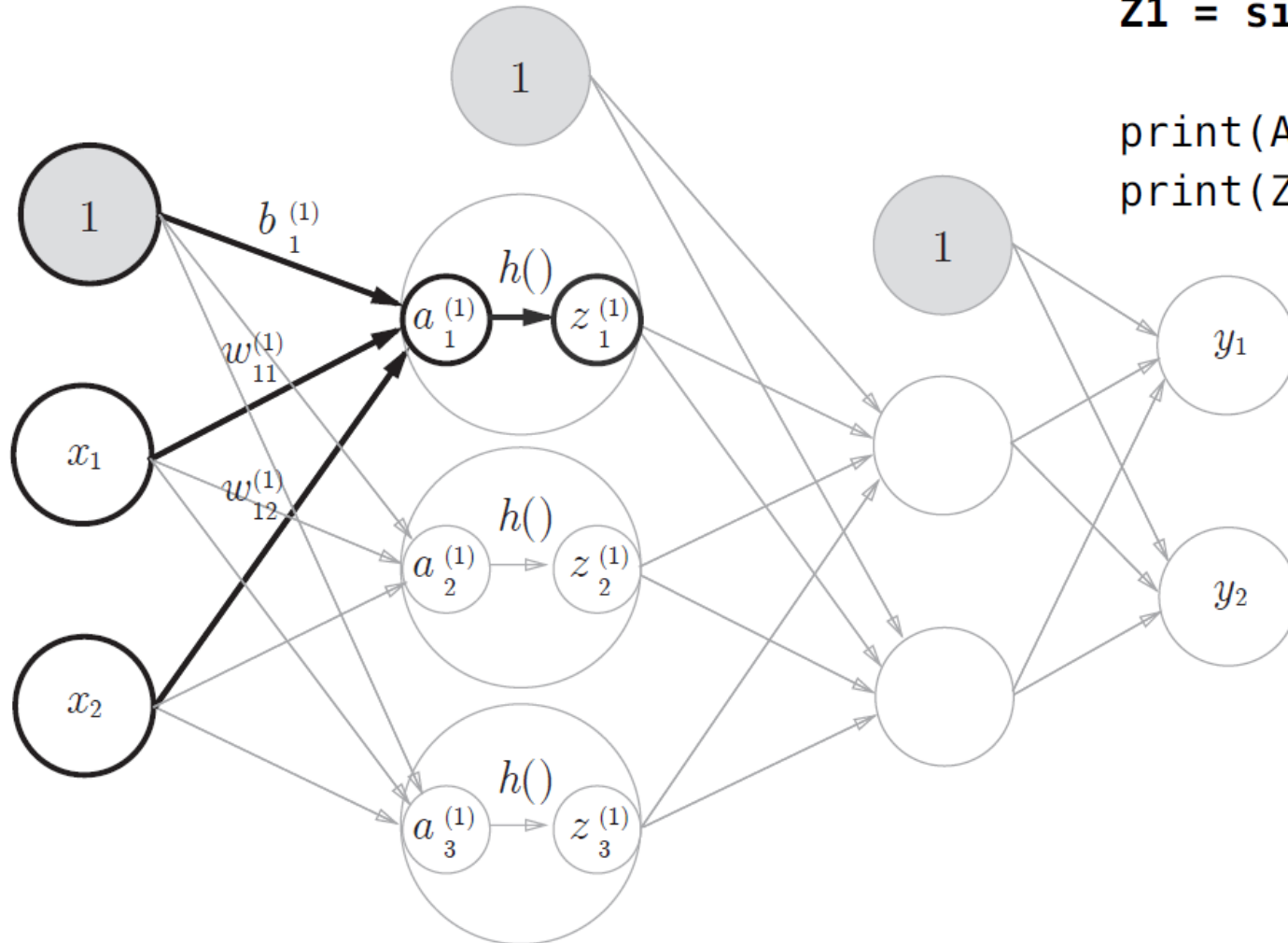


```
A1 = np.dot(X, W1) + B1
```

$$(1.0 \quad 0.5) \begin{pmatrix} 0.1 & 0.3 & 0.5 \\ 0.2 & 0.4 & 0.6 \end{pmatrix} + (0.1 \quad 0.2 \quad 0.3)$$

```
print(A1) # [0.3, 0.7, 1.1]
```

Process in 1st layer using activation function



Z1 = sigmoid(A1)

```
print(A1) # [0.3, 0.7, 1.1]
```

```
print(Z1) # [0.57444252, 0.66818777, 0.75026011]
```

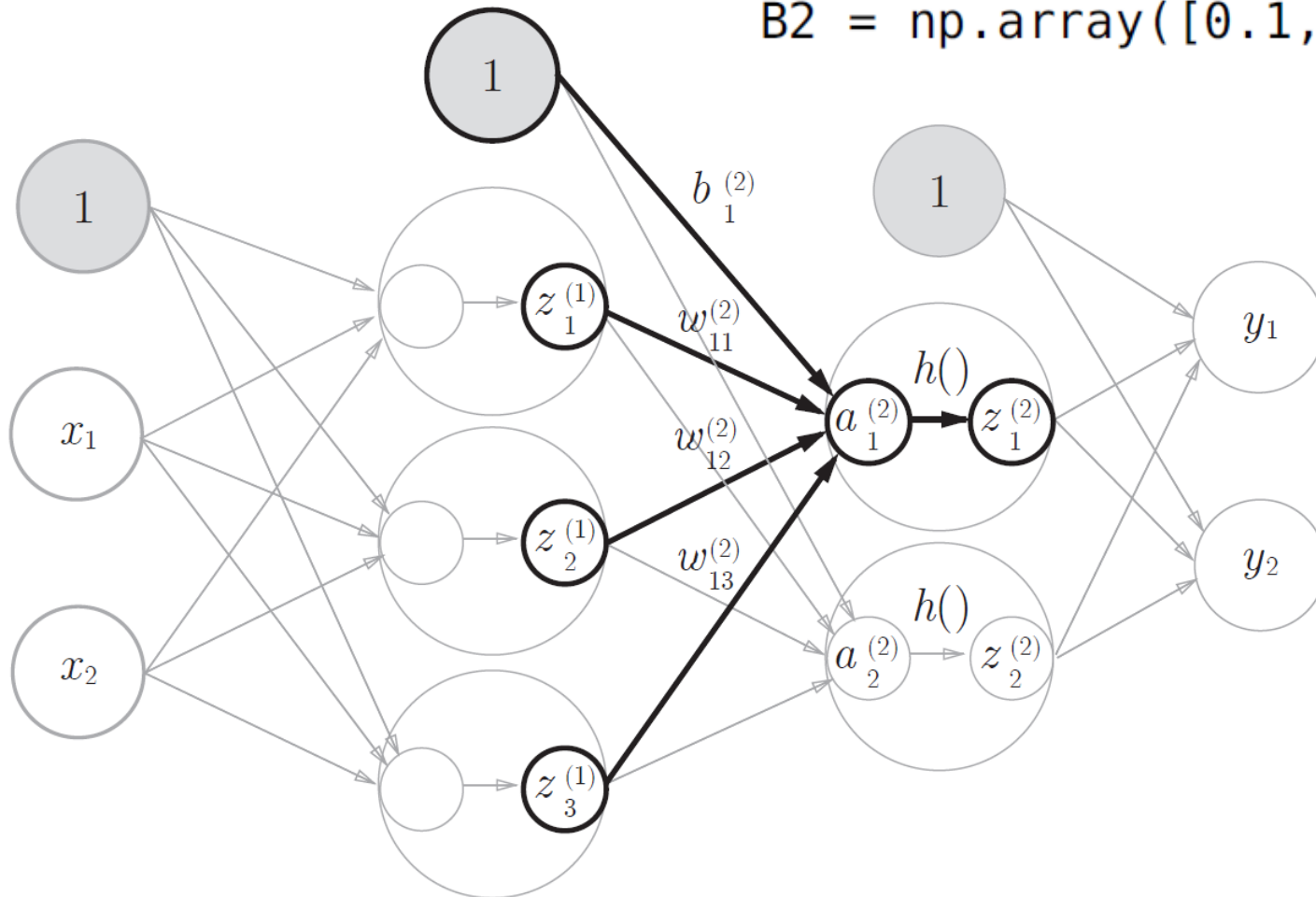
$$\left(\frac{1}{1 + e^{-0.3}}, \frac{1}{1 + e^{-0.7}}, \frac{1}{1 + e^{-1.1}} \right)$$

Sigmoid function

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

Signal transmission from 1st layer to 2nd layer

```
W2 = np.array([[0.1, 0.4], [0.2, 0.5], [0.3, 0.6]])  
B2 = np.array([0.1, 0.2])
```



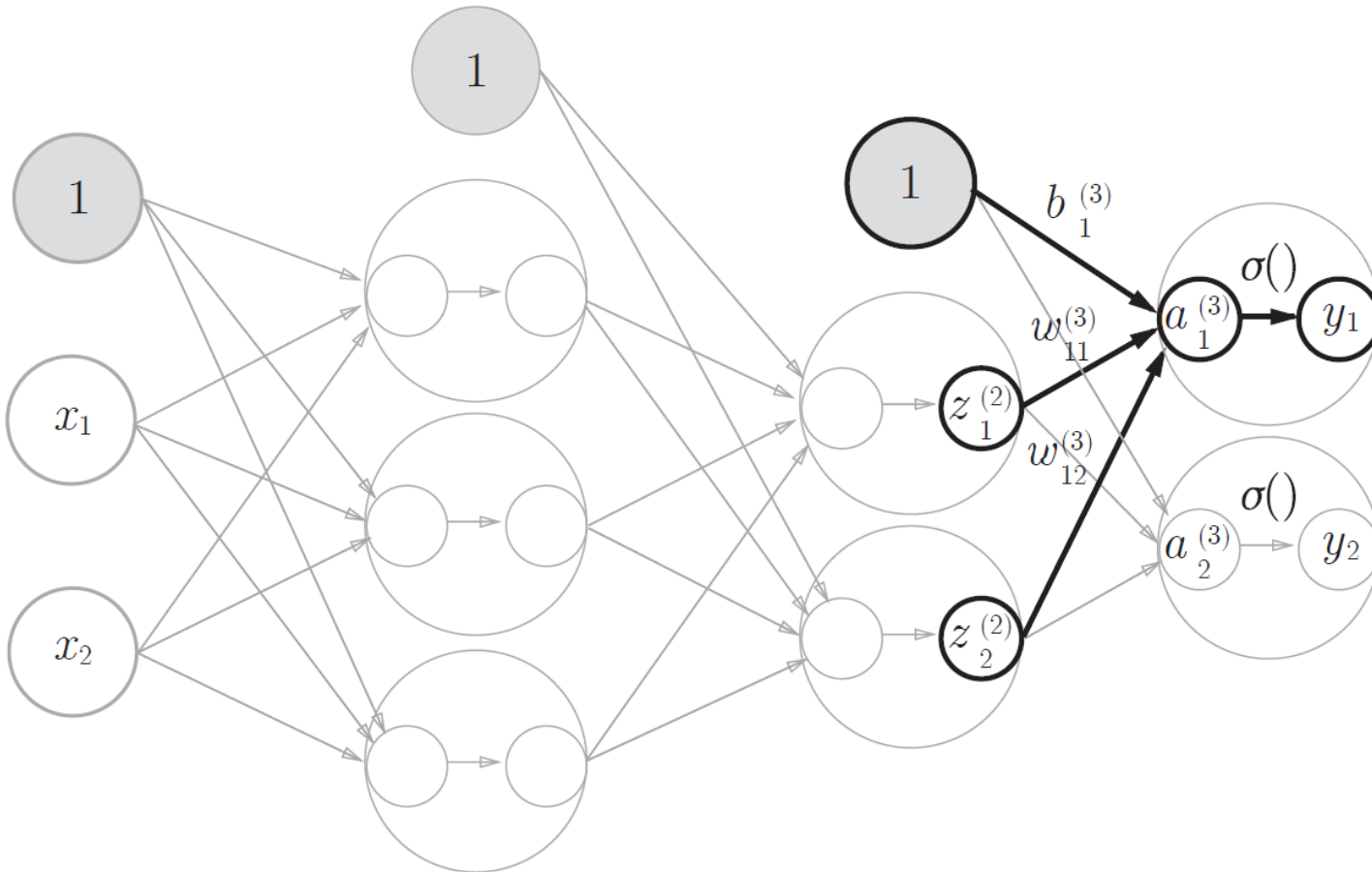
```
A2 = np.dot(Z1, W2) + B2  
Z2 = sigmoid(A2)
```

Signal transmission from 2nd layer to 3rd(Output) layer

```
def identity_function(x):  
    return x
```

```
W3 = np.array([[0.1, 0.3], [0.2, 0.4]])  
B3 = np.array([0.1, 0.2])
```

```
A3 = np.dot(Z2, W3) + B3  
Y = identity_function(A3)
```



Summary of implementation

```
def init_network():
    network = {}
    network['W1'] = np.array([[0.1, 0.3, 0.5], [0.2, 0.4, 0.6]])
    network['b1'] = np.array([0.1, 0.2, 0.3])
    network['W2'] = np.array([[0.1, 0.4], [0.2, 0.5], [0.3, 0.6]])
    network['b2'] = np.array([0.1, 0.2])
    network['W3'] = np.array([[0.1, 0.3], [0.2, 0.4]])
    network['b3'] = np.array([0.1, 0.2])

    return network
```

Set the weight and bias

```
def forward(network, x):
    W1, W2, W3 = network['W1'], network['W2'], network['W3']
    b1, b2, b3 = network['b1'], network['b2'], network['b3']

    a1 = np.dot(x, W1) + b1
    z1 = sigmoid(a1)
    a2 = np.dot(z1, W2) + b2
    z2 = sigmoid(a2)
    a3 = np.dot(z2, W3) + b3
    y = identity_function(a3)

    return y
```

Signal transmission
from input to output

```
network = init_network()
x = np.array([1.0, 0.5])
y = forward(network, x)
print(y) # [ 0.31682708  0.69627909]
```

Summary

- Multi-dimensional array is "Set of number"
- 2-dimensional array is called "matrix"
- Product of matrix make implementation of Neural Network efficient