

12주차(2/2)

# 심층 신경망 1

파이썬으로 배우는 기계학습

한동대학교  
김영섭 교수

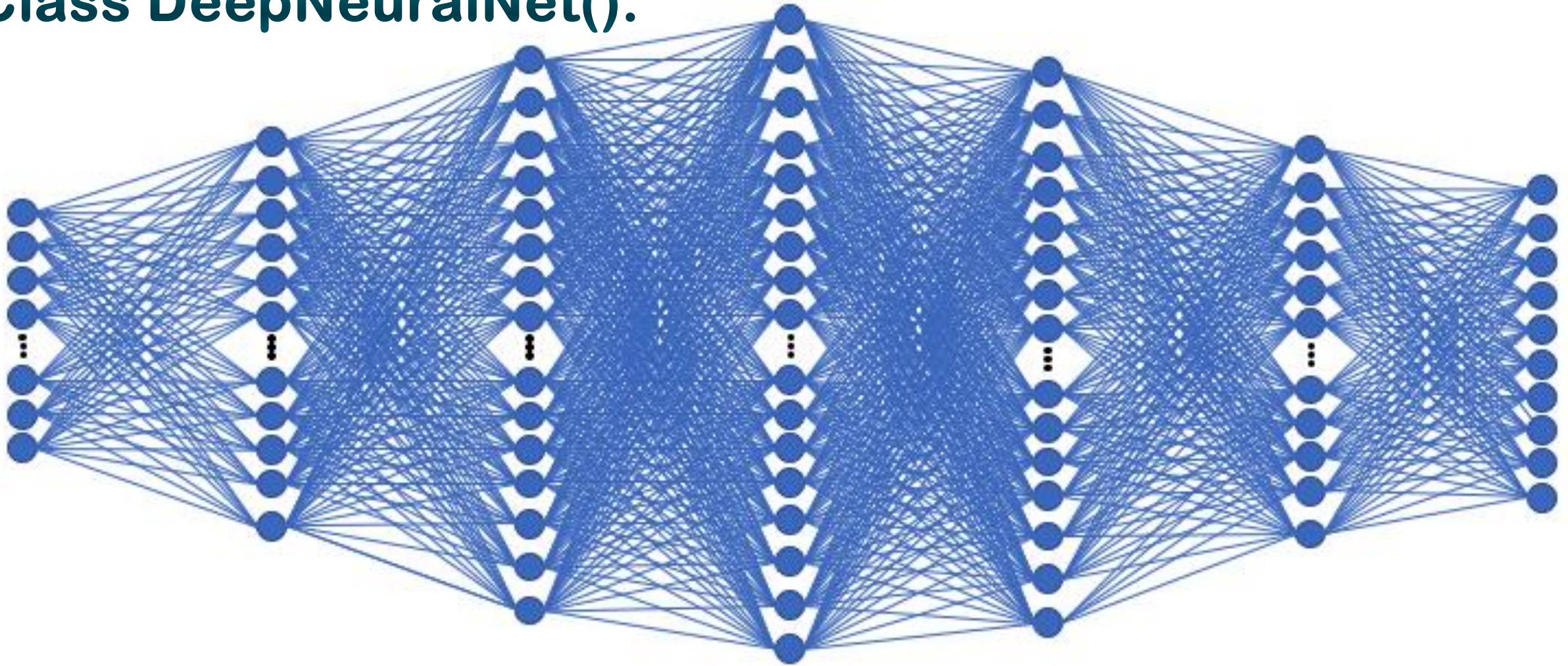
# 심층 신경망 1

---

- 학습 목표
  - 심층 신경망을 학습한다.
  - 은닉층의 개수에 따른 성능을 확인한다.
- 학습 내용
  - 심층 신경망 이해하기
  - 심층 신경망 구현하기
  - 심층 신경망 성능 확인하기

# 1. DNN 클래스

**Class DeepNeuralNet():**

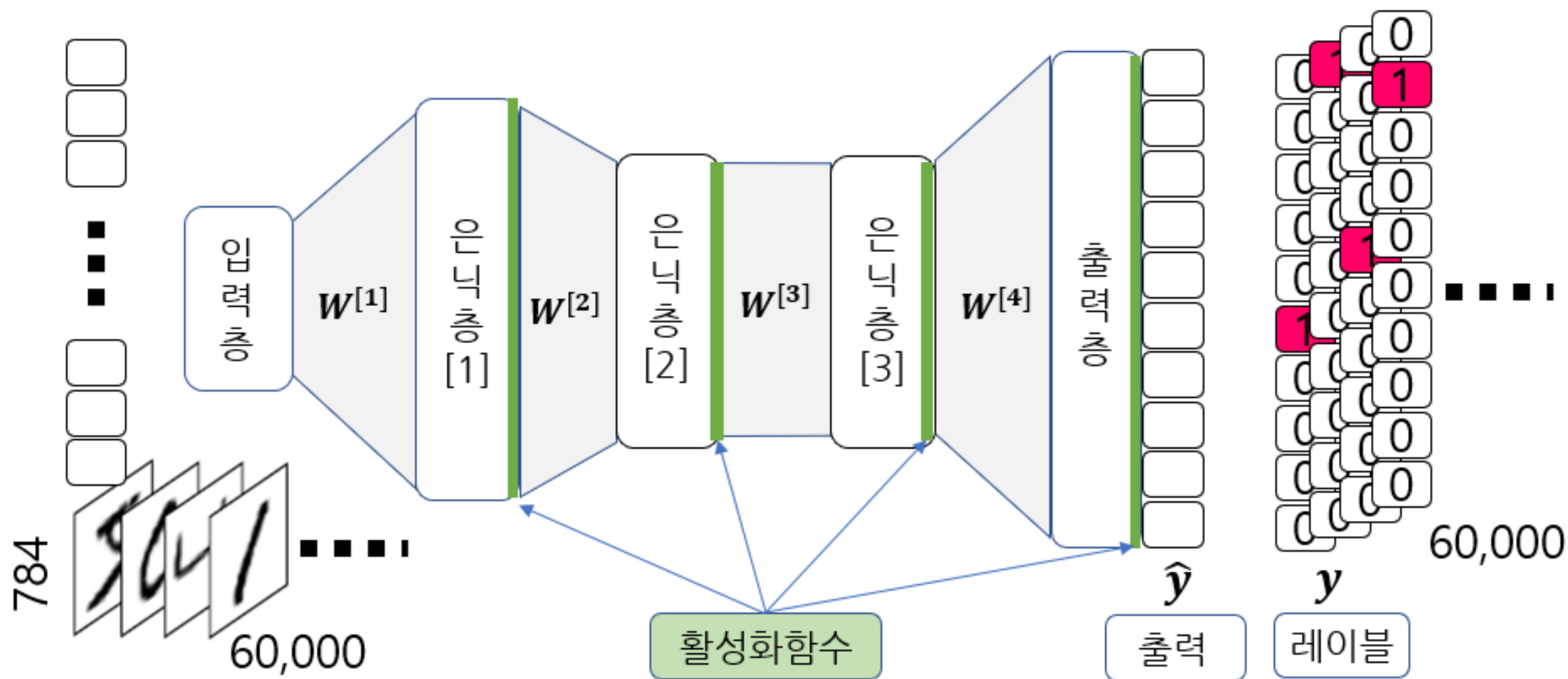




# 1. DNN 클래스 : 코드 개요

```
1 class DeepNeuralNet():
2     """ implements a deep neural net.
3         Users may specify any number of layers.
4         net_arch -- consists of a number of neurons in each layer
5     """
6     def __init__(self, net_arch, activate = None,
7                 eta = 1.0, epochs = 100, random_seed = 1):
8         pass
9
10    def forpass(self, A0):
11        pass
12
13    def backprop(self, Z, A, Y):
14        pass
15
16    def fit(self, X, y):
17        pass
18
19    def predict(self, X):
20        pass
21
22    def evaluate(self, Xtest, ytest):
23        pass
```

## 2. DNN 활성화 함수 : 구조



## 2. DNN 활성화 함수: tanh

$$\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} = \frac{1 - e^{-2x}}{1 + e^{-2x}}$$



```
1 def tanh(x):
2     return (1.0 - np.exp(-2 * x)) / (
3         1.0 + np.exp(-2 * x))
4
5 def tanh_d(x):
6     return (1 + tanh(x)) * (1 - tanh(x))
7
8 def sigmoid(x):
9     #x = np.clip(x, -500, 500)
10    return 1 / (1 + np.exp((-x)))
11
12 def sigmoid_d(x):
13    return sigmoid(x) * (1 - sigmoid(x))
14
15 def relu(x):
16    return np.maximum(x, 0)
17
18 def relu_d(x):
19    x[x<=0] = 0
20    x[x>0] = 1
21    return x
```

## 2. DNN 활성화 함수: tanh

$$\tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} = \frac{1 - e^{-2x}}{1 + e^{-2x}}$$

$$\begin{aligned}\frac{d}{dx} \tanh(x) &= 1 - \left( \frac{2}{1 + e^{-2x}} - 1 \right)^2 \\ &= 1 - \tanh^2(x)\end{aligned}$$



```
1 def tanh(x):
2     return (1.0 - np.exp(-2 * x)) / (
3         1.0 + np.exp(-2 * x))
4
5 def tanh_d(x):
6     return (1 + tanh(x)) * (1 - tanh(x))
7
8 def sigmoid(x):
9     #x = np.clip(x, -500, 500)
10    return 1 / (1 + np.exp((-x)))
11
12 def sigmoid_d(x):
13    return sigmoid(x) * (1 - sigmoid(x))
14
15 def relu(x):
16    return np.maximum(x, 0)
17
18 def relu_d(x):
19    x[x<=0] = 0
20    x[x>0] = 1
21    return x
```

## 2. DNN 활성화 함수: sigmoid

$$\sigma(x) = \frac{1}{1 + e^{-x}} = \frac{1}{1 + \frac{1}{e^x}}$$



```
1 def tanh(x):
2     return (1.0 - np.exp(-2 * x)) / (
3         1.0 + np.exp(-2 * x))
4
5 def tanh_d(x):
6     return (1 + tanh(x)) * (1 - tanh(x))
7
8 def sigmoid(x):
9     #x = np.clip(x, -500, 500)
10    return 1 / (1 + np.exp((-x)))
11
12 def sigmoid_d(x):
13    return sigmoid(x) * (1 - sigmoid(x))
14
15 def relu(x):
16    return np.maximum(x, 0)
17
18 def relu_d(x):
19    x[x<=0] = 0
20    x[x>0] = 1
21    return x
```



## 2. DNN 활성화 함수: sigmoid

$$\sigma(x) = \frac{1}{1 + e^{-x}} = \frac{1}{1 + \frac{1}{e^x}}$$

$$\begin{aligned} \frac{d}{dx} \sigma(x) &= \frac{1}{1 + e^{-x}} \left(1 - \frac{1}{1 + e^{-x}}\right) \\ &= \sigma(x)(1 - \sigma(x)) \end{aligned}$$



```
1 def tanh(x):
2     return (1.0 - np.exp(-2 * x)) / (
3         1.0 + np.exp(-2 * x))
4
5 def tanh_d(x):
6     return (1 + tanh(x)) * (1 - tanh(x))
7
8 def sigmoid(x):
9     #x = np.clip(x, -500, 500)
10    return 1 / (1 + np.exp((-x)))
11
12 def sigmoid_d(x):
13     return sigmoid(x) * (1 - sigmoid(x))
14
15 def relu(x):
16     return np.maximum(x, 0)
17
18 def relu_d(x):
19     x[x <= 0] = 0
20     x[x > 0] = 1
21     return x
```

## 2. DNN 활성화 함수: ReLU

$$\text{Relu}(x) = \begin{cases} x & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$



```
1 def tanh(x):
2     return (1.0 - np.exp(-2 * x)) / (
3         1.0 + np.exp(-2 * x))
4
5 def tanh_d(x):
6     return (1 + tanh(x)) * (1 - tanh(x))
7
8 def sigmoid(x):
9     #x = np.clip(x, -500, 500)
10    return 1 / (1 + np.exp((-x)))
11
12 def sigmoid_d(x):
13    return sigmoid(x) * (1 - sigmoid(x))
14
15 def relu(x):
16    return np.maximum(x, 0)
17
18 def relu_d(x):
19    x[x<=0] = 0
20    x[x>0] = 1
21    return x
```

## 2. DNN 활성화 함수: ReLU

$$\text{Relu}(x) = \begin{cases} x & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\frac{d}{dx} \text{Relu}(x) = \begin{cases} 1 & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$



```
1 def tanh(x):
2     return (1.0 - np.exp(-2 * x)) / (
3         1.0 + np.exp(-2 * x))
4
5 def tanh_d(x):
6     return (1 + tanh(x)) * (1 - tanh(x))
7
8 def sigmoid(x):
9     #x = np.clip(x, -500, 500)
10    return 1 / (1 + np.exp((-x)))
11
12 def sigmoid_d(x):
13    return sigmoid(x) * (1 - sigmoid(x))
14
15 def relu(x):
16    return np.maximum(x, 0)
17
18 def relu_d(x):
19    x[x<=0] = 0
20    x[x>0] = 1
21    return x
```

## 2. DNN 활성화 함수

```
1 def tanh(x):
2     return (1.0 - np.exp(-2 * x))/(
3         1.0 + np.exp(-2 * x))
4
5 def tanh_d(x):
6     return (1 + tanh(x)) * (1 - tanh(x))
7
8 def sigmoid(x):
9     #x = np.clip(x, -500, 500)
10    return 1 / (1 + np.exp((-x)))
11
12 def sigmoid_d(x):
13    return sigmoid(x) * (1 - sigmoid(x))
14
15 def relu(x):
16    return np.maximum(x, 0)
17
18 def relu_d(x):
19    x[x<=0] = 0
20    x[x>0] = 1
21    return x
```

### 3. DNN 구현: 생성자

```
def __init__(self, net_arch, activate = None,
              eta = 1.0, epochs = 100, random_seed = 1):
    self.eta = eta
    self.epochs = epochs
    self.net_arch = net_arch
    self.layers = len(net_arch)
    self.W = []

    self.g = [lambda x: sigmoid(x) for _ in range(self.layers)]
    self.g_prime = [lambda x: sigmoid_d(x) for _ in range(self.layers)]

    if activate is not None:
        for i, (g, g_prime) in enumerate(zip(activate[::2], activate[1::2])):
            self.g[i+1] = g
            self.g_prime[i+1] = g_prime

    np.random.seed(random_seed)
    self.W = [[None]] ## the first W0 is not used.
    for layer in range(self.layers - 1):
        w = 2 * np.random.rand(self.net_arch[layer+1],
                                self.net_arch[layer]) - 1

        self.W.append(w)
```



### 3. DNN 구현: 생성자

```
def __init__(self, net_arch, activate = None,
              eta = 1.0, epochs = 100, random_seed = 1):
    self.eta = eta
    self.epochs = epochs
    self.net_arch = net_arch
    self.layers = len(net_arch)
    self.W = []

    self.g = [lambda x: sigmoid(x) for _ in range(self.layers)]
    self.g_prime = [lambda x: sigmoid_d(x) for _ in range(self.layers)]

    if activate is not None:
        for i, (g, g_prime) in enumerate(zip(activate[::2], activate[1::2])):
            self.g[i+1] = g
            self.g_prime[i+1] = g_prime

    np.random.seed(random_seed)
    self.W = [[None]] ## the first W0 is not used.
    for layer in range(self.layers - 1):
        w = 2 * np.random.rand(self.net_arch[layer+1],
                                self.net_arch[layer]) - 1
        self.W.append(w)
```

(뒷층 노드 수, 앞층 노드 수)



### 3. DNN 구현: fit() 메소드

```
1  def fit(self, X, y):  
2      self.cost_ = []  
3      for epoch in range(self.epochs):  
4          Z, A = self.forpass(X)  
5          cost = self.backprop(Z, A, y)  
6          self.cost_.append(np.sqrt(  
7              np.sum(cost * cost)))  
8      return self
```

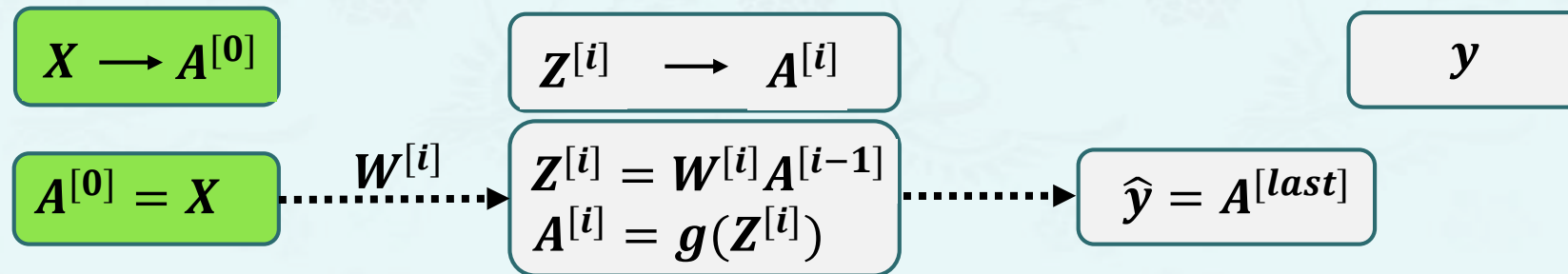
## 4. fit() 메소드 : 순전파

```
1 def fit(self, X, y):
2     self.cost_ = []
3     for epoch in range(self.epochs):
4         Z, A = self.forpass(X)
5         cost = self.backprop(Z, A, y)
6         self.cost_.append(np.sqrt(
7             np.sum(cost * cost)))
8     return self
```

## 4. fit() 메소드 : 순전파

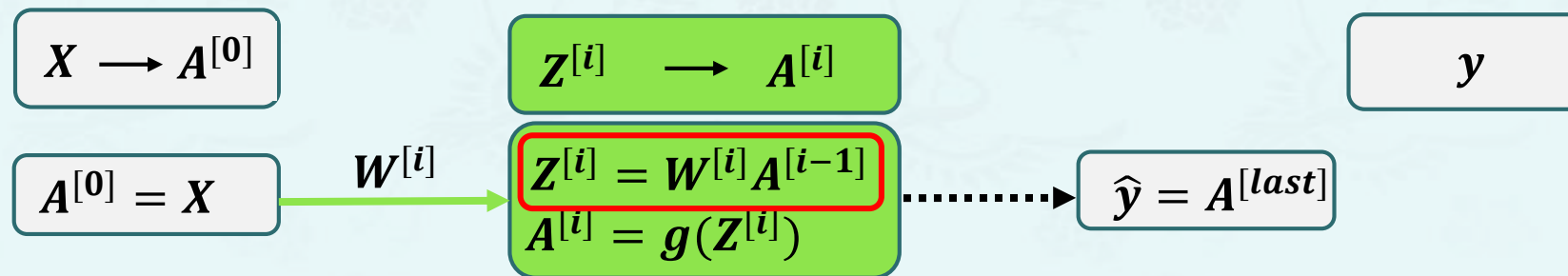
```
1 def forpass(self, A0):
2     Z = [[None]] # Z0 is not used.
3     A = []       # A0 = X0 is used.
4     A.append(A0)
5     for i in range(1, len(self.W)):
6         z = np.dot(self.W[i], A[i-1])
7         Z.append(z)
8         a = self.g[i](z)
9         A.append(a)
10    return Z, A
```

```
1 def fit(self, X, y):
2     self.cost_ = []
3     for epoch in range(self.epochs):
4         Z, A = self.forpass(X)
5         cost = self.backprop(Z, A, y)
6         self.cost_.append(np.sqrt(
7             np.sum(cost * cost)))
8     return self
```



## 4. fit() 메소드 : 순전파

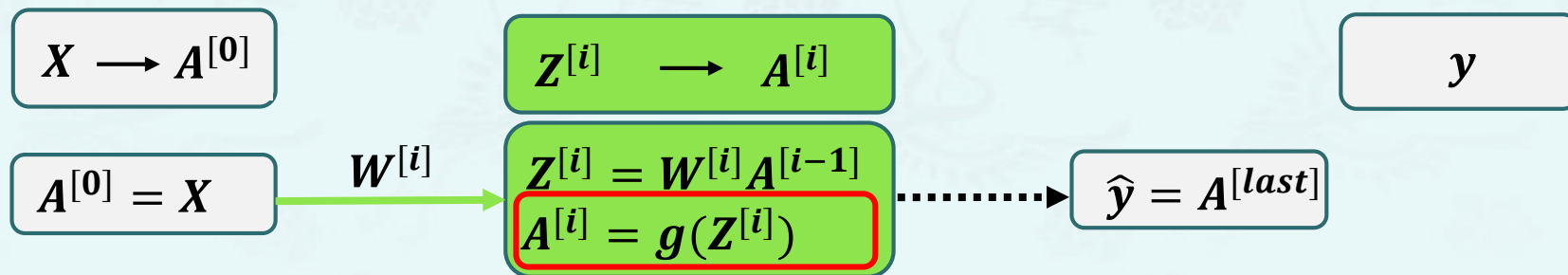
```
1 def forpass(self, A0):
2     Z = [[None]] # Z0 is not used.
3     A = []       # A0 = X0 is used.
4     A.append(A0)
5     for i in range(1, len(self.W)):
6         z = np.dot(self.W[i], A[i-1])
7         Z.append(z)
8         a = self.g[i](z)
9         A.append(a)
10    return Z, A
```





## 4. fit() 메소드 : 순전파

```
1 def forpass(self, A0):
2     Z = [[None]] # Z0 is not used.
3     A = []       # A0 = X0 is used.
4     A.append(A0)
5     for i in range(1, len(self.W)):
6         z = np.dot(self.W[i], A[i-1])
7         Z.append(z)
8         a = self.g[i](z)
9         A.append(a)
10    return Z, A
```



## 4. fit() 메소드 : 역전파

```
1 def fit(self, X, y):  
2     self.cost_ = []  
3     for epoch in range(self.epochs):  
4         Z, A = self.forpass(X)  
5         ➡ cost = self.backprop(Z, A, y)  
6         self.cost_.append(np.sqrt(  
7             np.sum(cost * cost)))  
8     return self
```

## 4. fit() 메소드 : 역전파

```

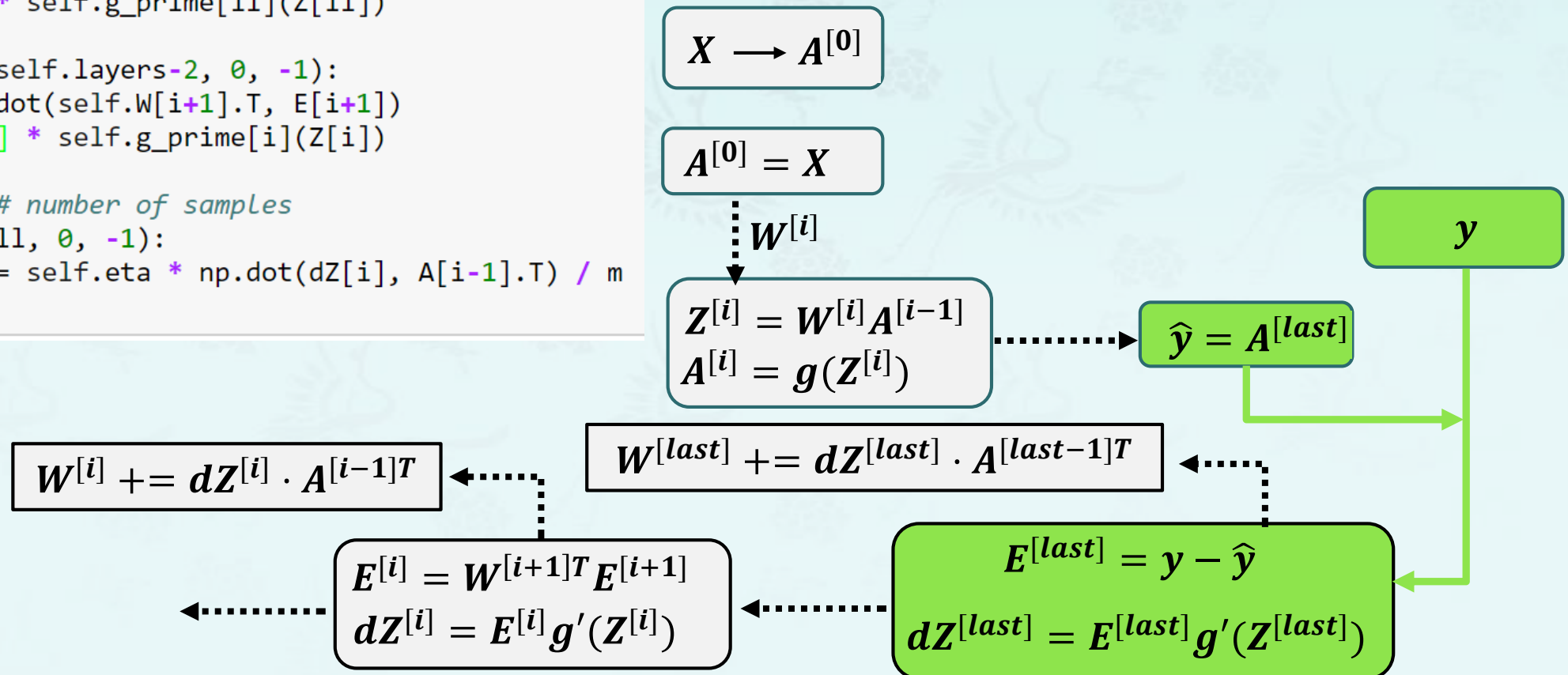
1 def backprop(self, Z, A, Y):
2     E = [None for x in range(self.layers)]
3     dZ = [None for x in range(self.layers)]
4
5     ll = self.layers - 1
6     error = Y - A[ll]
7     E[ll] = error
8     dZ[ll] = error * self.g_prime[ll](Z[ll])
9
10    for i in range(self.layers-2, 0, -1):
11        E[i] = np.dot(self.W[i+1].T, E[i+1])
12        dZ[i] = E[i] * self.g_prime[i](Z[i])
13
14    m = Y.shape[0] # number of samples
15    for i in range(ll, 0, -1):
16        self.W[i] += self.eta * np.dot(dZ[i], A[i-1].T) / m
17    return error

```

```

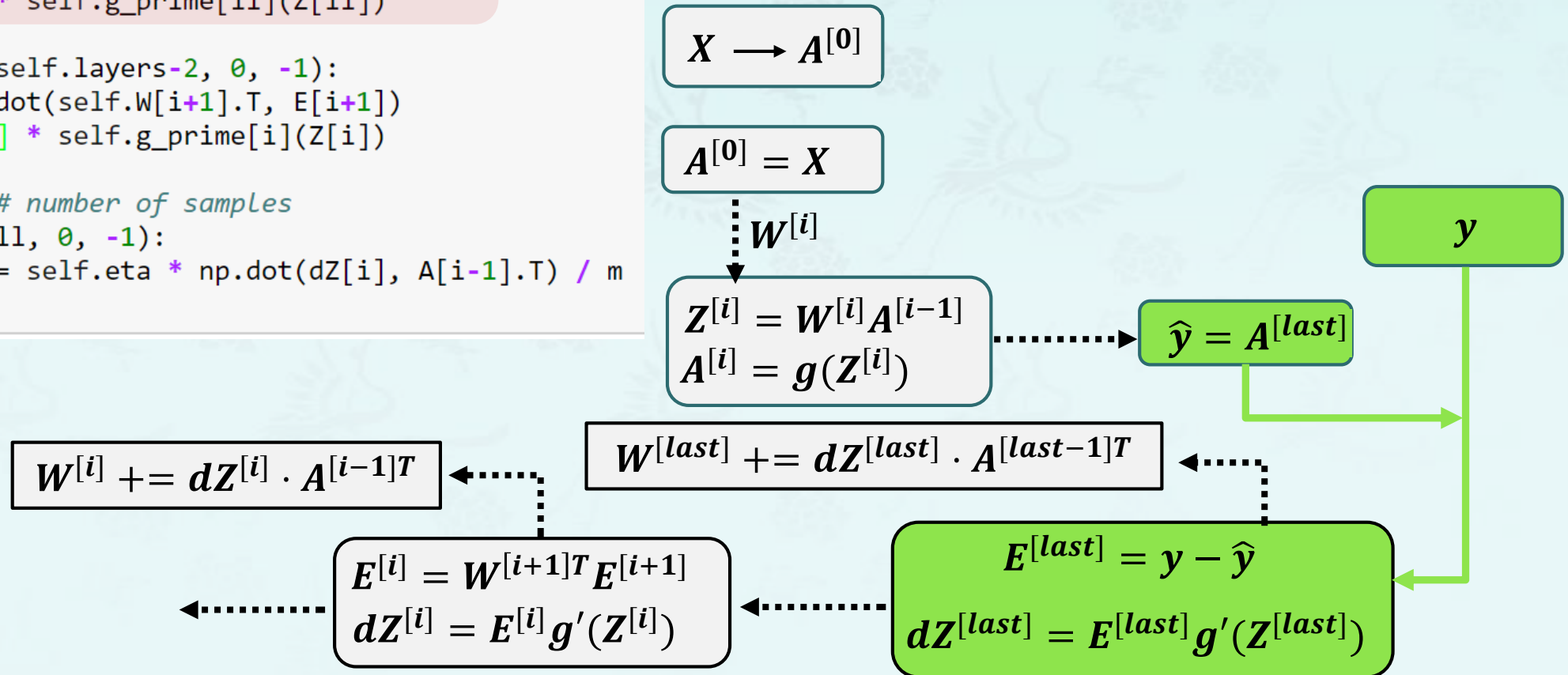
1 def fit(self, X, y):
2     self.cost_ = []
3     for epoch in range(self.epochs):
4         Z, A = self.forpass(X)
5         cost = self.backprop(Z, A, y)
6         self.cost_.append(np.sqrt(np.sum(cost * cost)))
7     return self

```



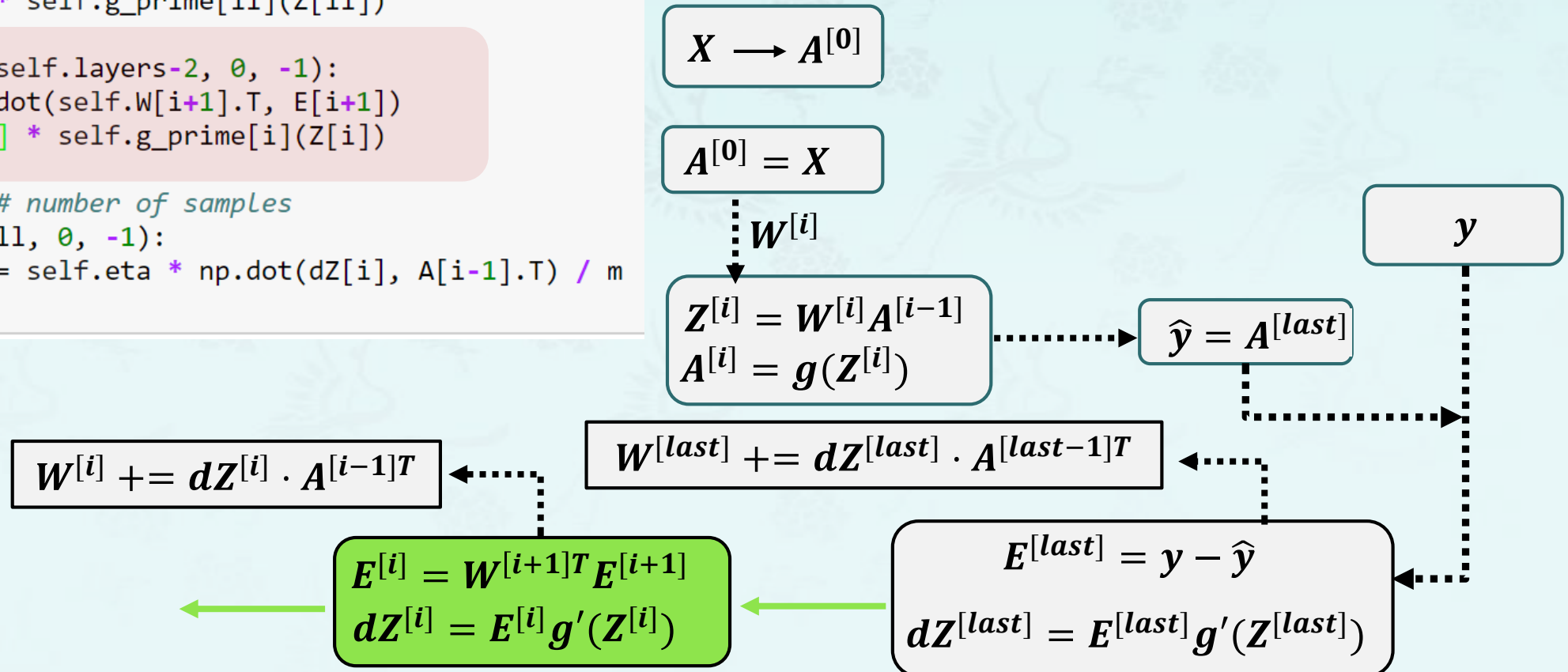
## 4. fit() 메소드 : 역전파

```
1 def backprop(self, Z, A, Y):
2     E = [None for x in range(self.layers)]
3     dZ = [None for x in range(self.layers)]
4
5     ll = self.layers - 1
6     error = Y - A[ll]
7     E[ll] = error
8     dZ[ll] = error * self.g_prime[ll](Z[ll])
9
10    for i in range(self.layers-2, 0, -1):
11        E[i] = np.dot(self.W[i+1].T, E[i+1])
12        dZ[i] = E[i] * self.g_prime[i](Z[i])
13
14    m = Y.shape[0] # number of samples
15    for i in range(ll, 0, -1):
16        self.W[i] += self.eta * np.dot(dZ[i], A[i-1].T) / m
17    return error
```



## 4. fit() 메소드 : 역전파

```
1 def backprop(self, Z, A, Y):
2     E = [None for x in range(self.layers)]
3     dZ = [None for x in range(self.layers)]
4
5     ll = self.layers - 1
6     error = Y - A[ll]
7     E[ll] = error
8     dZ[ll] = error * self.g_prime[ll](Z[ll])
9
10    for i in range(self.layers-2, 0, -1):
11        E[i] = np.dot(self.W[i+1].T, E[i+1])
12        dZ[i] = E[i] * self.g_prime[i](Z[i])
13
14    m = Y.shape[0] # number of samples
15    for i in range(ll, 0, -1):
16        self.W[i] += self.eta * np.dot(dZ[i], A[i-1].T) / m
17    return error
```



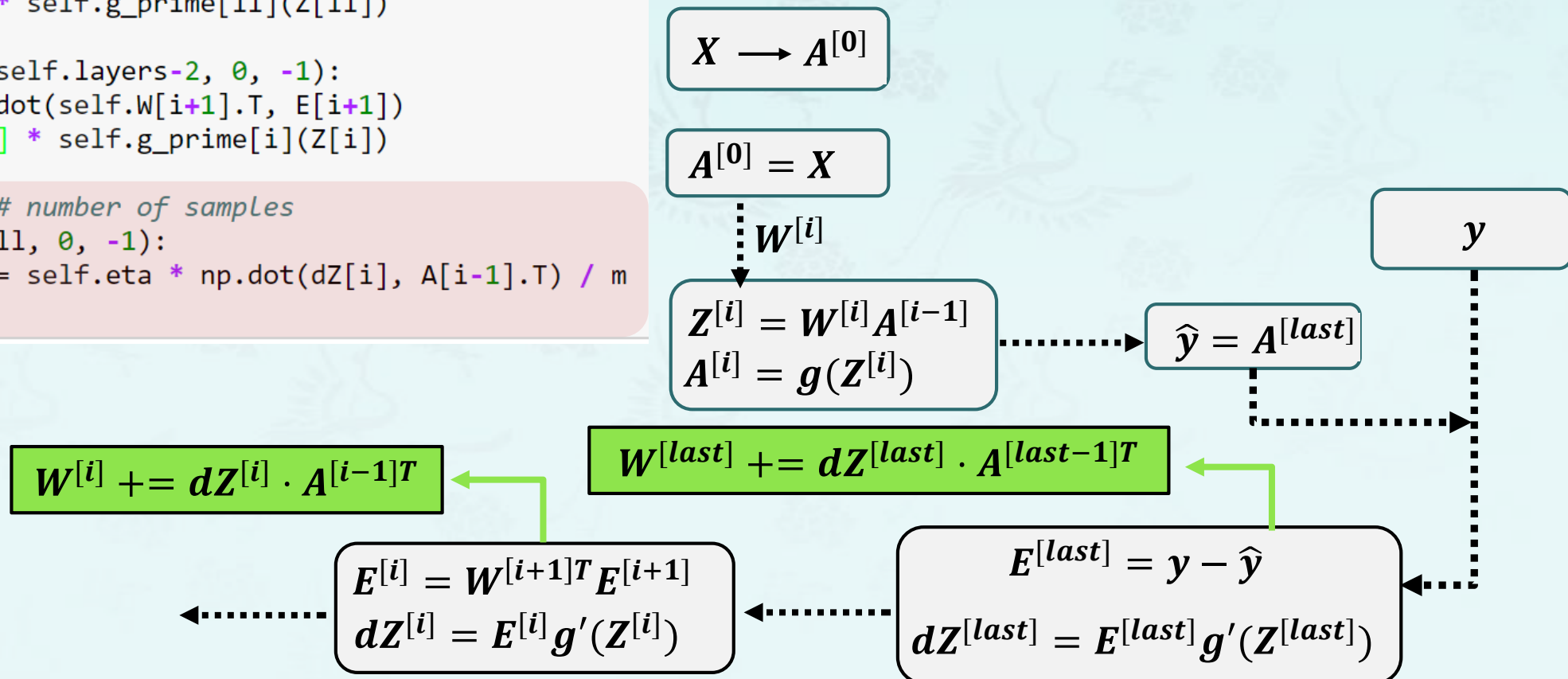


## 4. fit() 메소드 : 역전파


```

1 def backprop(self, Z, A, Y):
2     E = [None for x in range(self.layers)]
3     dZ = [None for x in range(self.layers)]
4
5     ll = self.layers - 1
6     error = Y - A[ll]
7     E[ll] = error
8     dZ[ll] = error * self.g_prime[ll](Z[ll])
9
10    for i in range(self.layers-2, 0, -1):
11        E[i] = np.dot(self.W[i+1].T, E[i+1])
12        dZ[i] = E[i] * self.g_prime[i](Z[i])
13
14    m = Y.shape[0] # number of samples
15    for i in range(ll, 0, -1):
16        self.W[i] += self.eta * np.dot(dZ[i], A[i-1].T) / m
17    return error

```



## 4. fit() 메소드 : 오차 계산

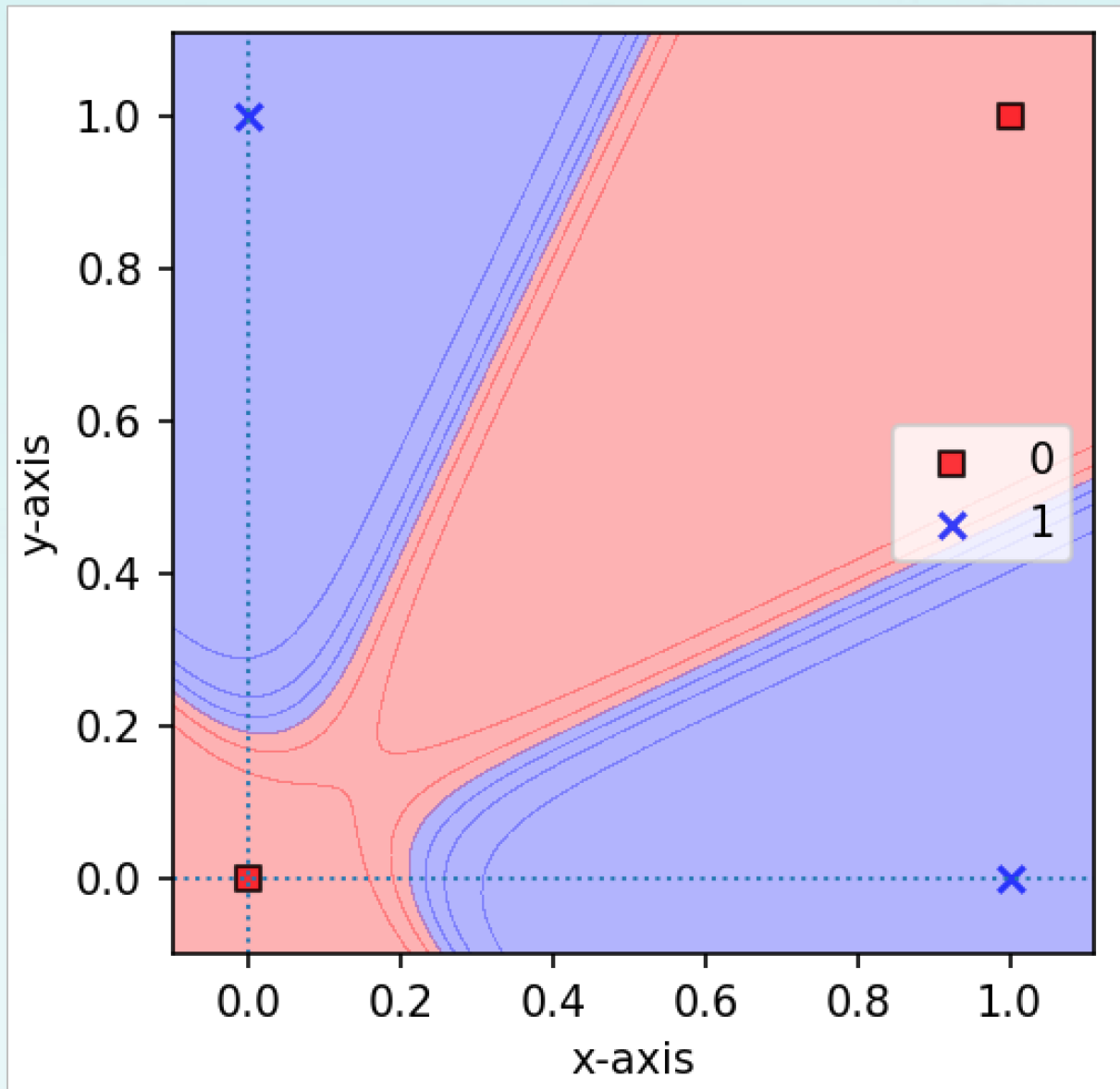
```
1 def fit(self, X, y):  
2     self.cost_ = []  
3     for epoch in range(self.epochs):  
4         Z, A = self.forpass(X)  
5         cost = self.backprop(Z, A, y)  
6          self.cost_.append(np.sqrt(  
7             np.sum(cost * cost)))  
8     return self
```

## 5. DNN 학습 결과: XOR 테스트

```
1 import joy
2 X = np.array([ [0, 0, 1, 1], [0, 1, 0, 1] ])
3 y = np.array([0, 1, 1, 0])
4 dnn = DeepNeuralNet([2, 4, 2, 1], eta = 0.5, epochs = 5000).fit(X, y)
5
6 joy.plot_decision_regions(X.T, y, dnn)
7 plt.xlabel('x-axis')
8 plt.ylabel('y-axis')
9 plt.legend(loc='best')
10 plt.show()
```

[입력층, 은닉층, 은닉층, 출력층]의 노드 수

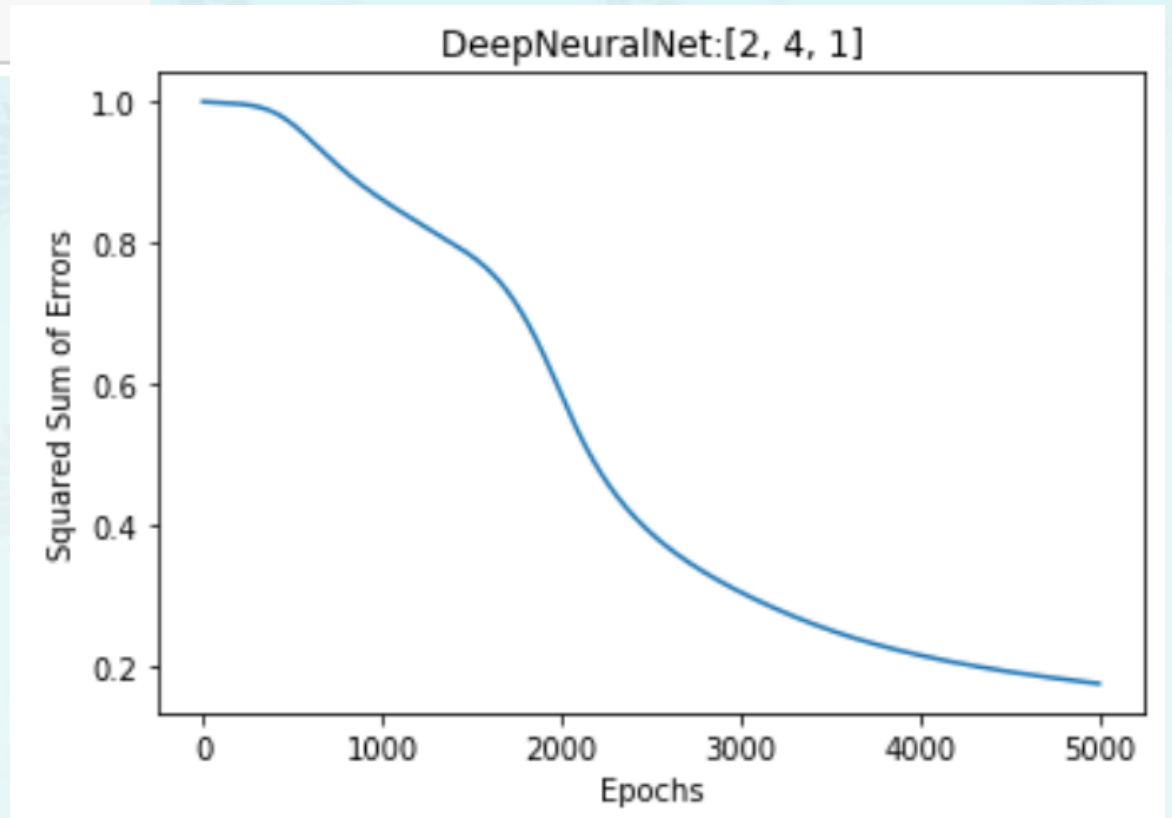
## 5. DNN 학습 결과: XOR 테스트



## 5. DNN 학습 결과: XOR 테스트

```
dnn = DeepNeuralNet([2, 4, 1],  
                    eta = 0.5, epochs = 5000).fit(X, y)  
plt.plot(range(len(dnn.cost_)), dnn.cost_)  
plt.xlabel('Epochs')  
plt.ylabel('Squared Sum of Errors')  
plt.title('DeepNeuralNet:{}'.format(dnn.net_arch))  
plt.show()
```

은닉층 1개





## 5. DNN 학습 결과: 비교를 위한 학습 코딩

```
dnn1 = DeepNeuralNet([2,4,1], eta = 0.5, epochs = 5000).fit(X, y)
g = [sigmoid, sigmoid_d, sigmoid, sigmoid_d, sigmoid, sigmoid_d]
dnn2 = DeepNeuralNet([2,4,2,1], activate=g, eta = 0.5, epochs = 5000).fit(X, y)
plt.plot(range(len(dnn1.cost_)), dnn1.cost_, label='{}'.format(dnn1.net_arch))
plt.plot(range(len(dnn2.cost_)), dnn2.cost_, label='{}'.format(dnn2.net_arch))
plt.title('DeepNeuralNet:{} vs {}'.format(dnn1.net_arch, dnn2.net_arch))
plt.xlabel('Epochs')
plt.ylabel('Squared Sum of Errors')
plt.legend(loc='best')
plt.show()
```

## 5. DNN 학습 결과: 비교를 위한 학습 코딩

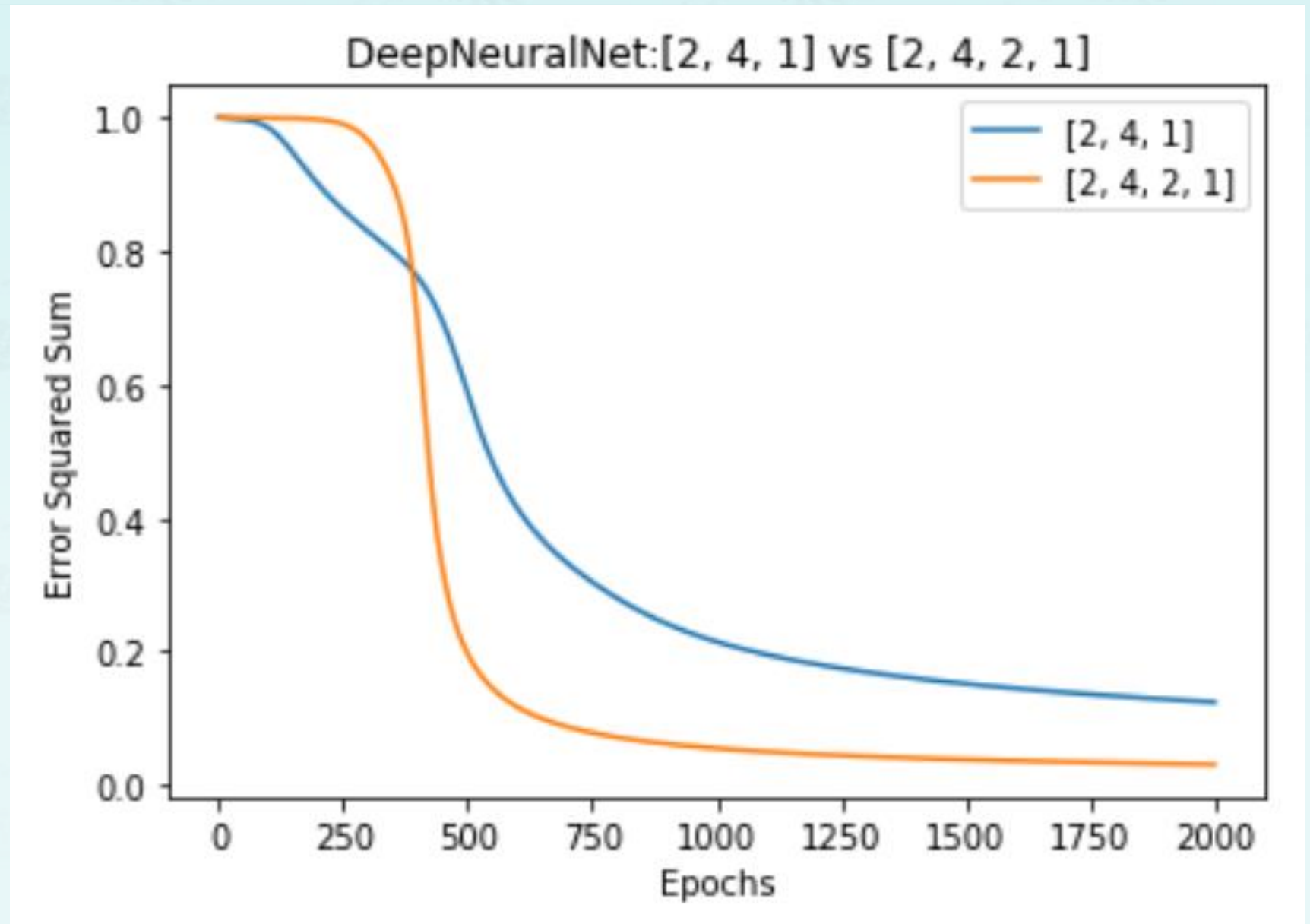
```
dnn1 = DeepNeuralNet([2,4,1], eta = 0.5, epochs = 5000).fit(X, y)
g = [sigmoid, sigmoid_d, sigmoid, sigmoid_d, sigmoid, sigmoid_d]
dnn2 = DeepNeuralNet([2,4,2,1], activate=g, eta = 0.5, epochs = 5000).fit(X, y)
plt.plot(range(len(dnn1.cost_)), dnn1.cost_, label='{}'.format(dnn1.net_arch))
plt.plot(range(len(dnn2.cost_)), dnn2.cost_, label='{}'.format(dnn2.net_arch))
plt.title('DeepNeuralNet:{} vs {}'.format(dnn1.net_arch, dnn2.net_arch))
plt.xlabel('Epochs')
plt.ylabel('Squared Sum of Errors')
plt.legend(loc='best')
plt.show()
```

은닉층 2개

## 5. DNN 학습 결과: 비교를 위한 학습 코딩

```
dnn1 = DeepNeuralNet([2,4,1], eta = 0.5, epochs = 5000).fit(X, y)
g = [sigmoid, sigmoid_d, sigmoid, sigmoid_d, sigmoid, sigmoid_d]
dnn2 = DeepNeuralNet([2,4,2,1], activate=g, eta = 0.5, epochs = 5000).fit(X, y)
plt.plot(range(len(dnn1.cost_)), dnn1.cost_, label='{}'.format(dnn1.net_arch))
plt.plot(range(len(dnn2.cost_)), dnn2.cost_, label='{}'.format(dnn2.net_arch))
plt.title('DeepNeuralNet:{} vs {}'.format(dnn1.net_arch, dnn2.net_arch))
plt.xlabel('Epochs')
plt.ylabel('Squared Sum of Errors')
plt.legend(loc='best')
plt.show()
```

## 5. DNN 학습 결과: 시각화 결과



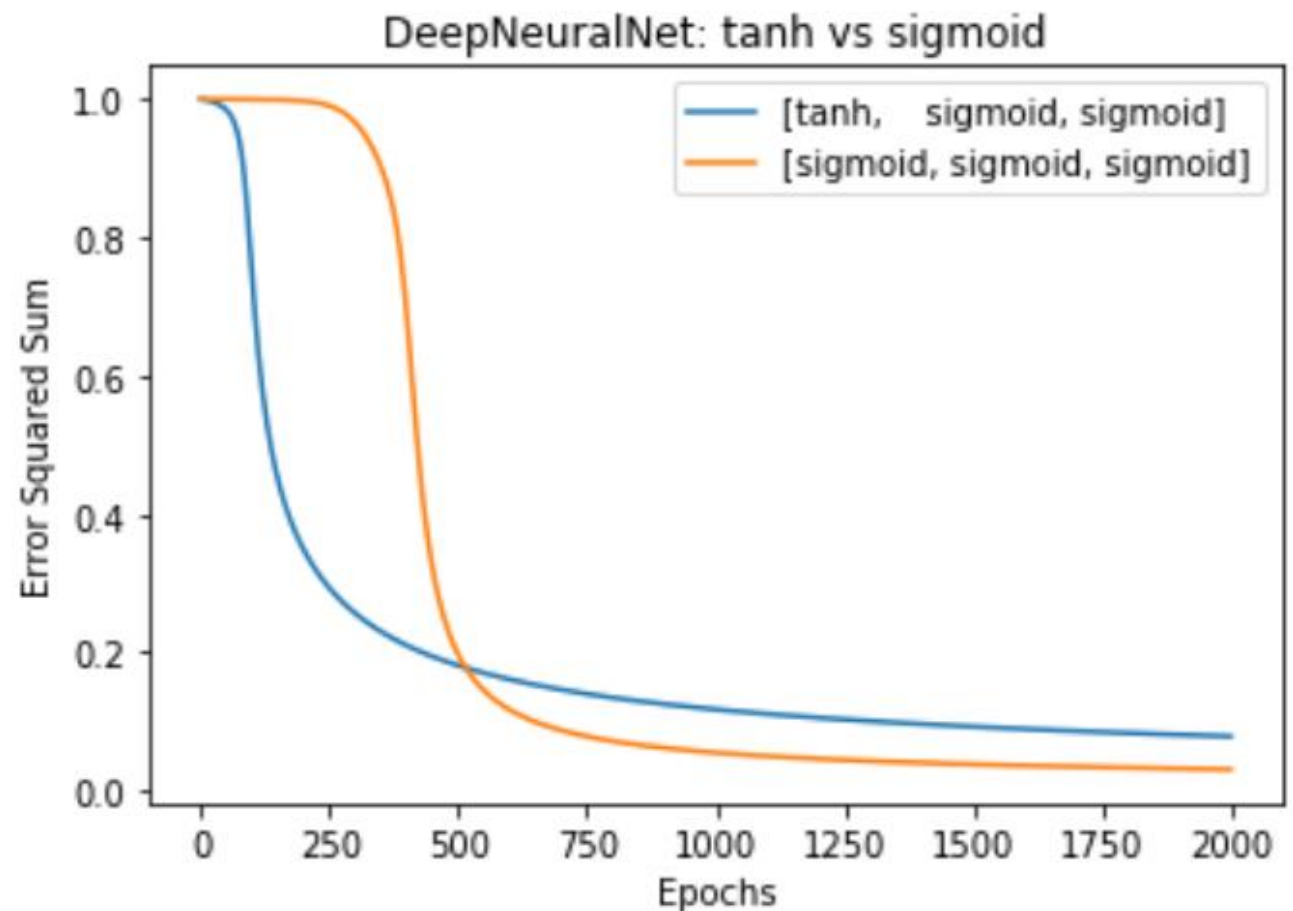
## 6. DNN 학습: 각 층별로 활성화 함수 지정

```
g1 = [tanh, tanh_d, sigmoid, sigmoid_d, sigmoid, sigmoid_d]
dnn1 = DeepNeuralNet([2,4,2,1], activate=g1, eta = 0.5, epochs = 5000).fit(X, y)
g2 = [sigmoid, sigmoid_d, sigmoid, sigmoid_d, sigmoid, sigmoid_d]
dnn2 = DeepNeuralNet([2,4,2,1], activate=g2, eta = 0.5, epochs = 5000).fit(X, y)
plt.plot(range(len(dnn1.cost_)), dnn1.cost_, label='[tanh, sigmoid, sigmoid]')
plt.plot(range(len(dnn2.cost_)), dnn2.cost_, label='[sigmoid, sigmoid, sigmoid]')
plt.title('DeepNeuralNet: tanh vs sigmoid')
plt.xlabel('Epochs')
plt.ylabel('Squared Sum of Errors')
plt.legend(loc='best')
plt.show()
```



## 6. DNN 학습: 각 층별로 활성화 함수 지정

```
g1 = [tanh, tanh_d, sigmoid, sigmoid_d, sigmoid, sigmoid_d]
dnn1 = DeepNeuralNet([2,4,2,1], activate=g1, eta = 0.5, epochs = 5000).fit(X, y)
g2 = [sigmoid, sigmoid_d, sigmoid, sigmoid_d, sigmoid, sigmoid_d]
dnn2 = DeepNeuralNet([2,4,2,1], activate=g2, eta = 0.5, epochs = 5000).fit(X, y)
plt.plot(range(len(dnn1.cost_)), dnn1.cost_)
plt.plot(range(len(dnn2.cost_)), dnn2.cost_)
plt.title('DeepNeuralNet: tanh vs sigmoid')
plt.xlabel('Epochs')
plt.ylabel('Squared Sum of Errors')
plt.legend(loc='best')
plt.show()
```





# 심층 신경망 1

---

- 학습 정리
  - 심층 신경망인 **DeepNeuralNet** 클래스 구현
  - 심층 신경망의 은닉층 갯수에 따른 성능 확인