11주차(1/3)

다층 신경망의 행렬 모델링

파이썬으로배우는기계학습

한동대학교 김영섭교수

다층 신경망의 행렬 모델링

- 학습 목표
 - 미분의 연쇄법칙을 학습한다.
 - 오차함수의 행렬 표기에서 미분하는 방법을 학습한다.
 - 다층 인공 신경망의 행렬 모델을 학습한다.
- 학습 내용
 - 미분의 연쇄법칙
 - 오차함수의 행렬 미분
 - 다층 인공 신경망 행렬 모델

$$F(x) = f(g(x))$$

$$F(x) = f(g(x))$$

$$F'(x) = f'(g(x))g'(x)$$

$$F(x) = f(g(x))$$

$$F'(x) = f'(g(x))g'(x)$$

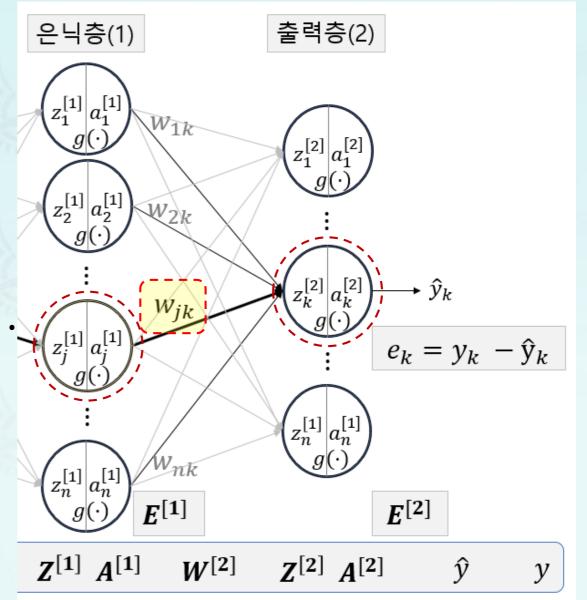
$$y = f(u), u = g(x)$$
 일 때, $\frac{dy}{dx} = \frac{dy}{du} \cdot \frac{du}{dx}$ (1)번 (2)번

$$\frac{\partial E}{\partial w_{jk}} = -(y_k - \hat{y}_k) \cdot \frac{\partial}{\partial w_{jk}} g(z_k)$$

$$= -(y_k - \hat{y}_k) \cdot g'(z_k) \frac{\partial z_k}{\partial w_{jk}}$$

$$= -(y_k - \hat{y}_k) \cdot g'(z_k) \frac{\partial}{\partial w_{jk}} (\sum_j w_{jk} \cdot y_{jk})$$

$$= -(y_k - \hat{y}_k) \cdot g'(z_k) \cdot a_j$$



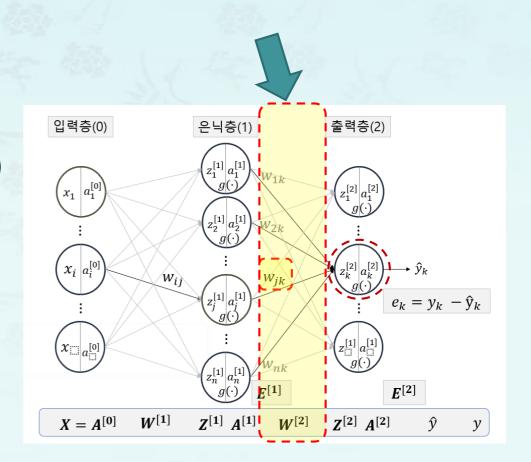
$$\frac{\partial E}{\partial w_{jk}} = -(y_k - \hat{y}_k) \cdot \frac{\partial}{\partial w_{jk}} g(z_k)$$

$$= -(y_k - \hat{y}_k) \cdot g'(z_k) \frac{\partial z_k}{\partial w_{jk}}$$

$$= -(y_k - \hat{y}_k) \cdot g'(z_k) \frac{\partial}{\partial w_{jk}} (\sum_j w_{jk} \cdot a_j)$$

$$= -(y_k - \hat{y}_k) \cdot g'(z_k) \cdot a_j$$

$$\frac{\partial E}{\partial W^{[2]}} = -E^{[2]} \cdot g'(Z^{[2]}) \cdot A^{[1]T}$$



$$W^{[2]} := W^{[2]} - \alpha \Delta W^{[2]}$$
$$= W^{[2]} - \alpha \frac{\partial E}{\partial W^{[2]}}$$

• 2단계
$$w_{jk}^{[2]} \coloneqq w_{jk}^{[2]} - \alpha \Delta w_{jk}^{[2]}$$

$$= w_{jk}^{[2]} - \alpha \frac{\partial E}{\partial w_{jk}^{[2]}}$$
• 3단계

$$\frac{\partial E}{\partial w_{ik}} = -(y_k - \hat{y}_k) \cdot \frac{\partial}{\partial w_{ik}} g(z_k)$$

• 4단계
$$\Delta w_{jk}^{[2]} = \frac{\partial E}{\partial w_{jk}} = -(y_k - \hat{y}_k) \cdot g'(z_k) \cdot a_j$$

$$\Delta W^{[2]} = \frac{\partial E}{\partial W^{[2]}} = -E^{[2]} \cdot g'(Z^{[2]}) \cdot A^{[1]T}$$

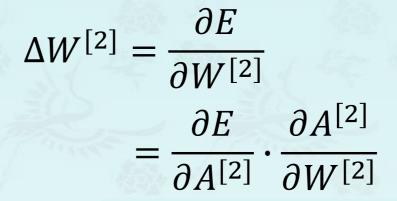
$$W^{[2]} := W^{[2]} - \alpha \Delta W^{[2]}$$

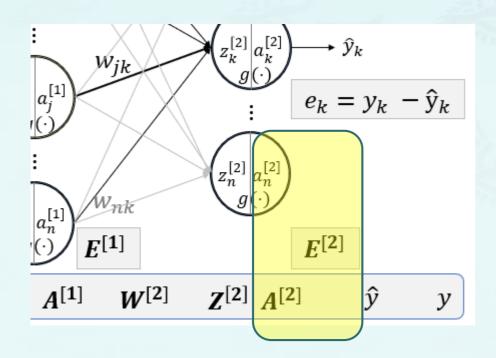
$$= W^{[2]} - \alpha \frac{\partial E}{\partial W^{[2]}}$$

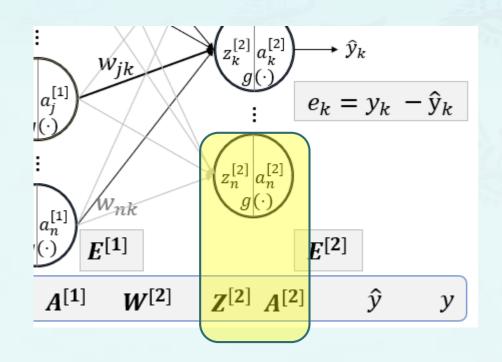
$$= W^{[2]} + E^{[2]} \cdot g'(Z^{[2]}) \cdot A^{[1]T}$$

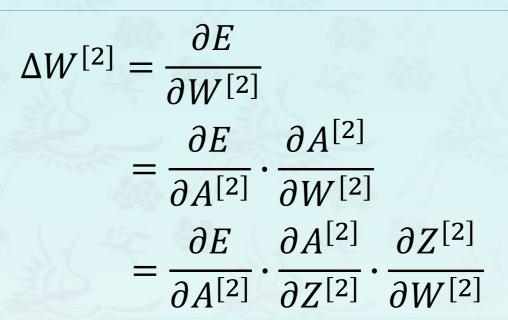
$$W^{[2]} := W^{[2]} - \alpha \Delta W^{[2]}$$
$$= W^{[2]} - \alpha \frac{\partial E}{\partial W^{[2]}}$$

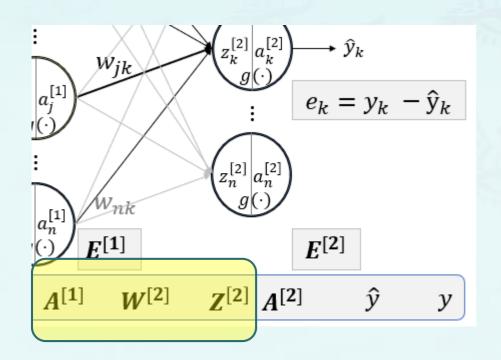
$$\Delta W^{[2]} = \frac{\partial E}{\partial W^{[2]}}$$

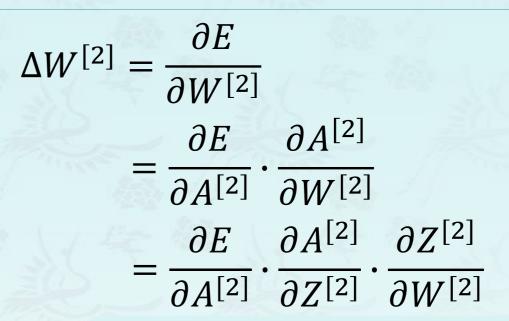








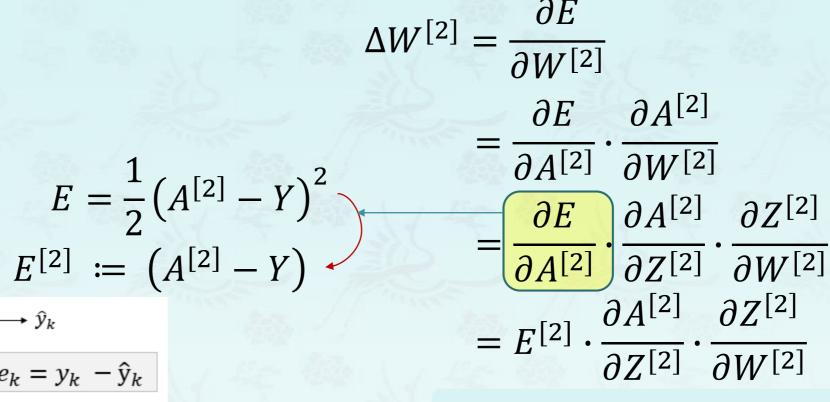


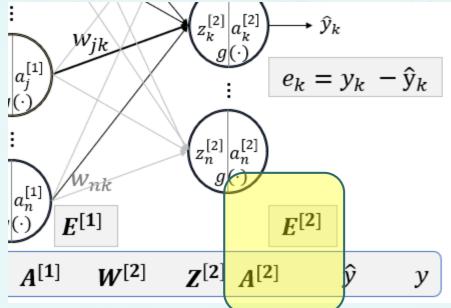


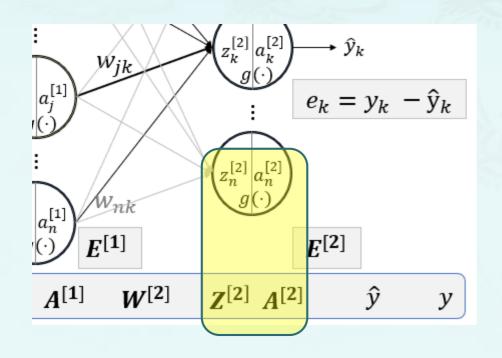
$$\Delta W^{[2]} = \frac{\partial E}{\partial W^{[2]}}$$

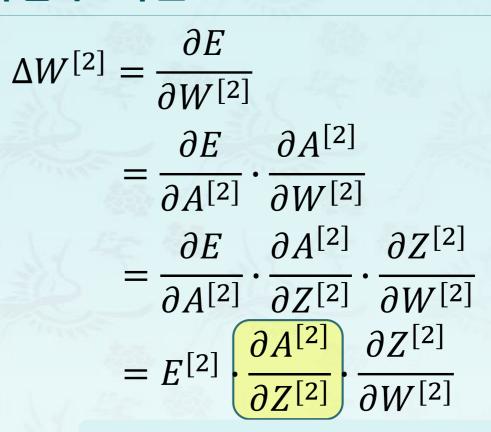
$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial W^{[2]}}$$

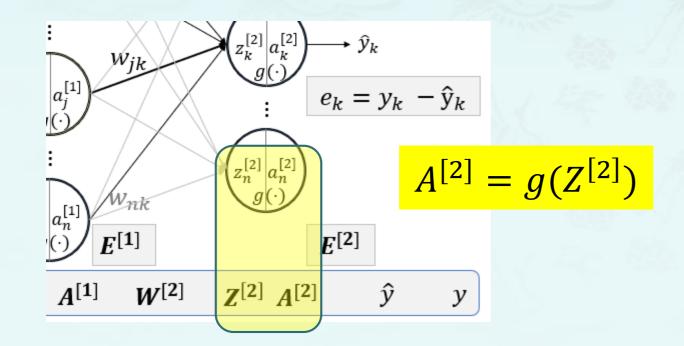
$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$











$$\Delta W^{[2]} = \frac{\partial E}{\partial W^{[2]}}$$

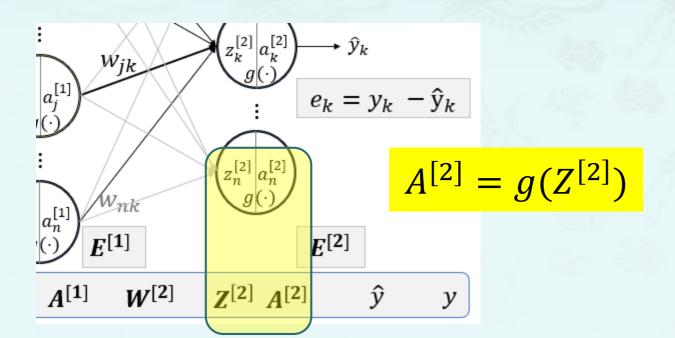
$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial W^{[2]}}$$

$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

$$= E^{[2]} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

$$= E^{[2]} \cdot g'(Z^{[2]}) \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

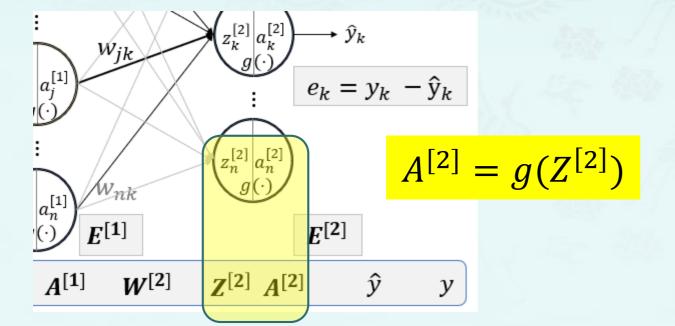
$$E^{[2]} \cdot \sigma(Z^{[2]}) (1 - \sigma(Z^{[2]})) \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$



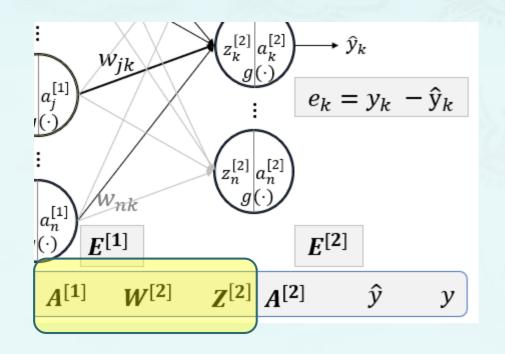
$$\begin{aligned} [2] &= \frac{\partial E}{\partial W^{[2]}} \\ &= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial W^{[2]}} \\ &= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}} \\ &= E^{[2]} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}} \\ &= E^{[2]} \cdot \frac{\partial Z^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}} \end{aligned}$$

$$E^{[2]} \cdot \sigma(Z^{[2]}) (1 - \sigma(Z^{[2]})) \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

$$= E^{[2]} \cdot A^{[2]} (1 - A^{[2]}) \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$



$$\begin{aligned} z^{[2]} &= \frac{\partial E}{\partial W^{[2]}} \\ &= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial W^{[2]}} \\ &= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}} \\ &= E^{[2]} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}} \\ &= E^{[2]} \cdot \frac{\partial Z^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}} \end{aligned}$$



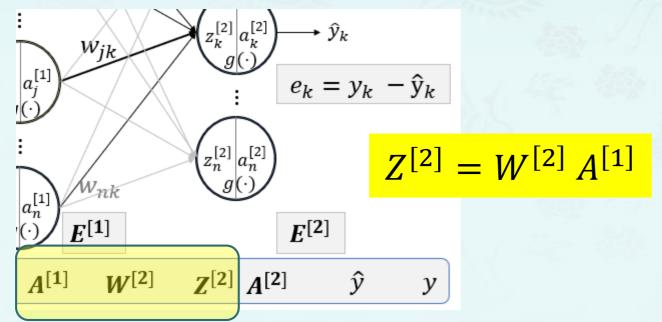
$$\Delta W^{[2]} = \frac{\partial E}{\partial W^{[2]}}$$

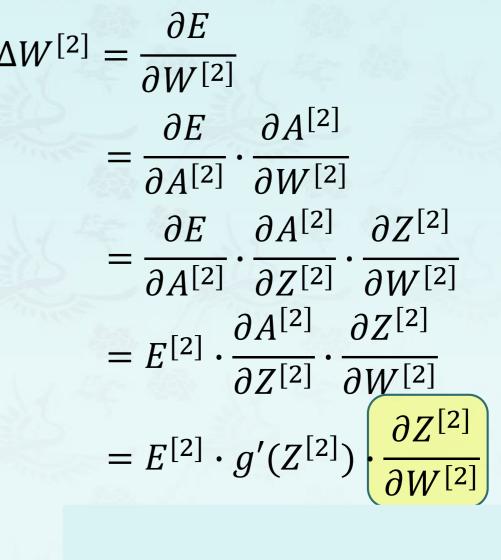
$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial W^{[2]}}$$

$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

$$= E^{[2]} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

$$= E^{[2]} \cdot g'(Z^{[2]}) \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$





$$\Delta W^{[2]} = \frac{\partial E}{\partial W^{[2]}}$$

$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial W^{[2]}}$$

$$= \frac{\partial E}{\partial A^{[2]}} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

$$= E^{[2]} \cdot \frac{\partial A^{[2]}}{\partial Z^{[2]}} \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

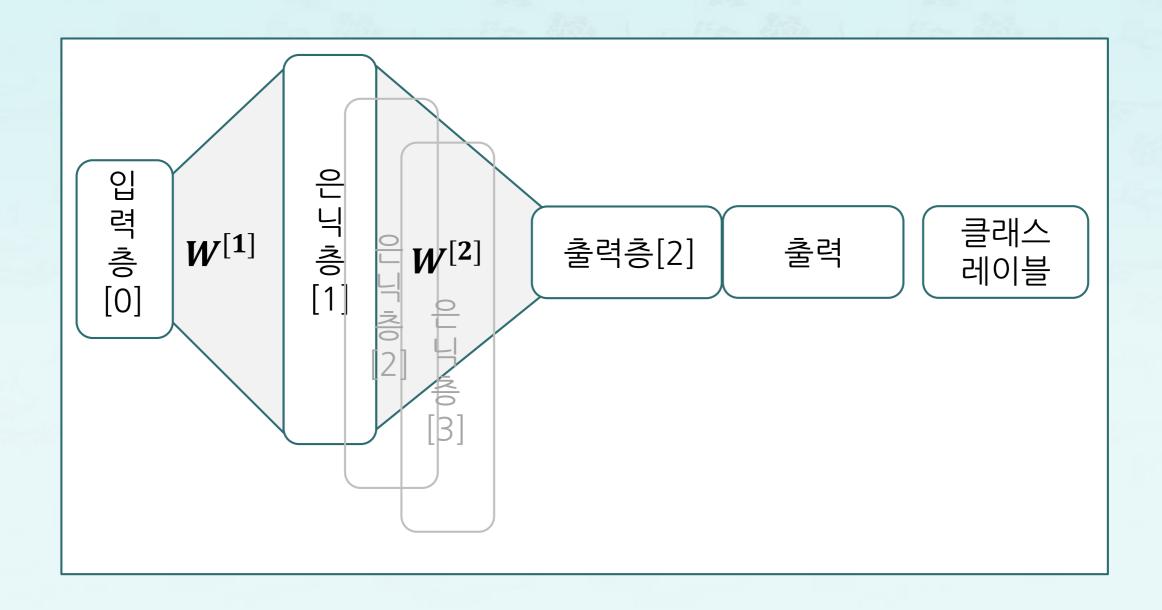
$$= E^{[2]} \cdot g'(Z^{[2]}) \cdot \frac{\partial Z^{[2]}}{\partial W^{[2]}}$$

$$= E^{[2]} \cdot g'(Z^{[2]}) \cdot A^{[1] \cdot T}$$

다층 인공신경망 행렬 모델:

■ 다층 신경망의 구조

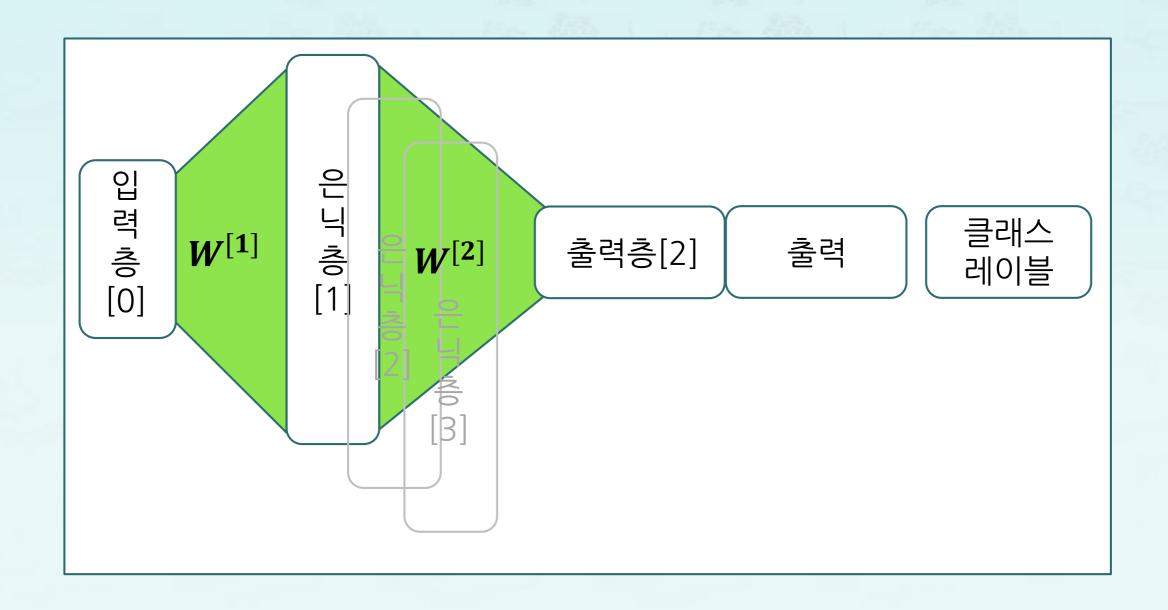
다층 인공신경망 행렬 모델:



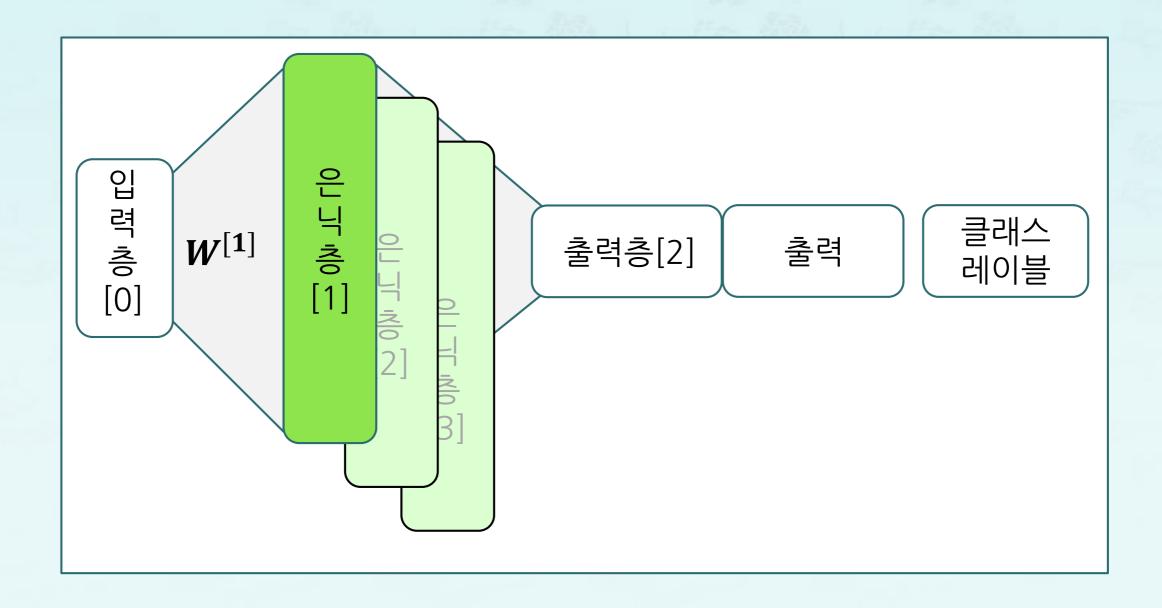
다층 인공신경망 행렬 모델: 입력층, 출력층



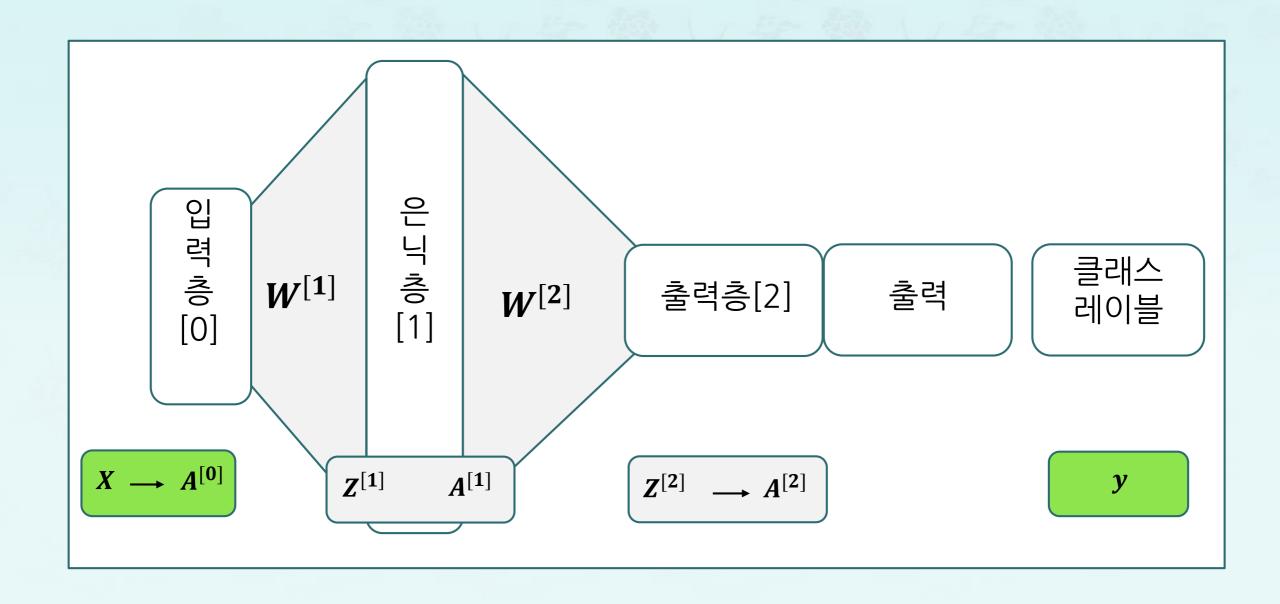
다층 인공신경망 행렬 모델: 가중치



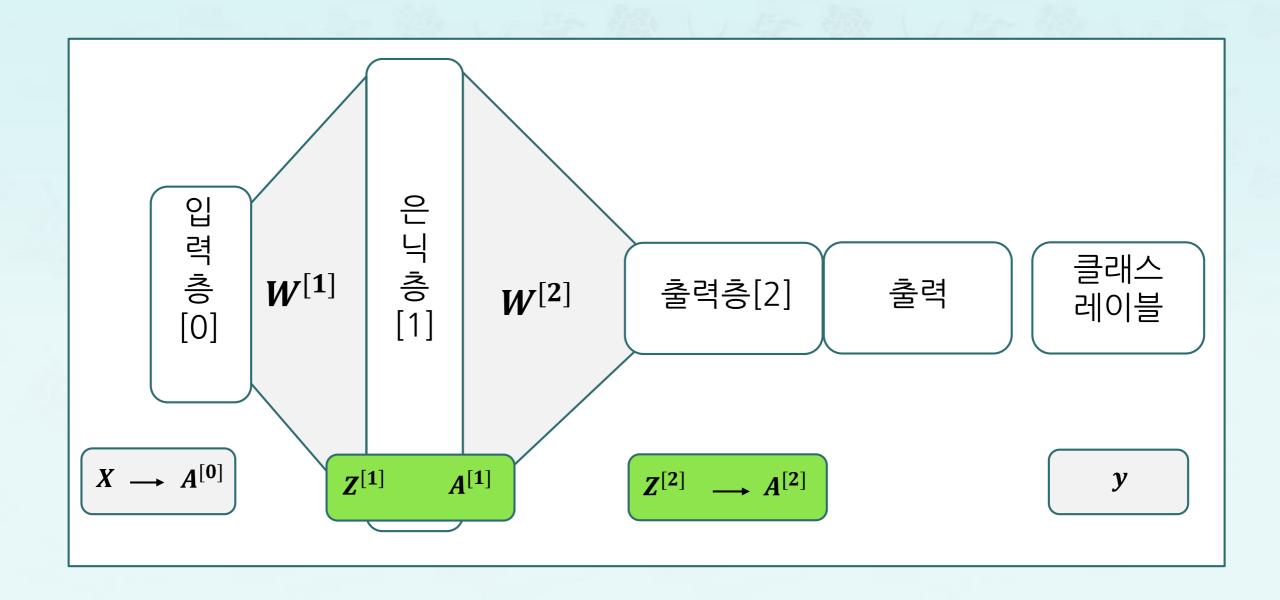
다층 인공신경망 행렬 모델: 은닉층

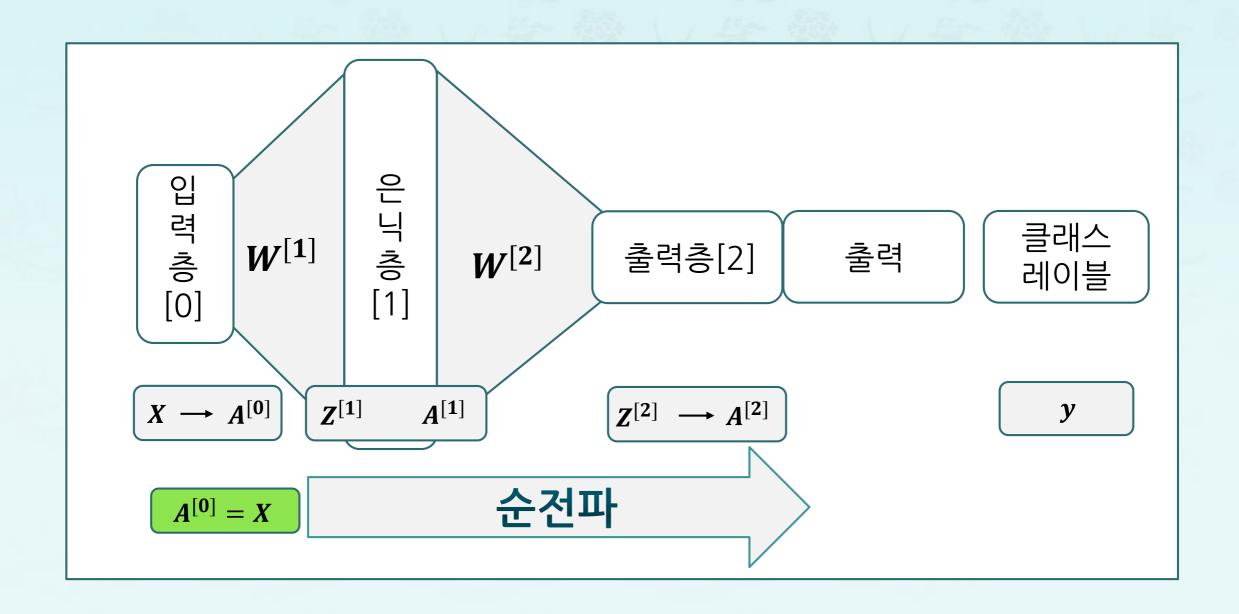


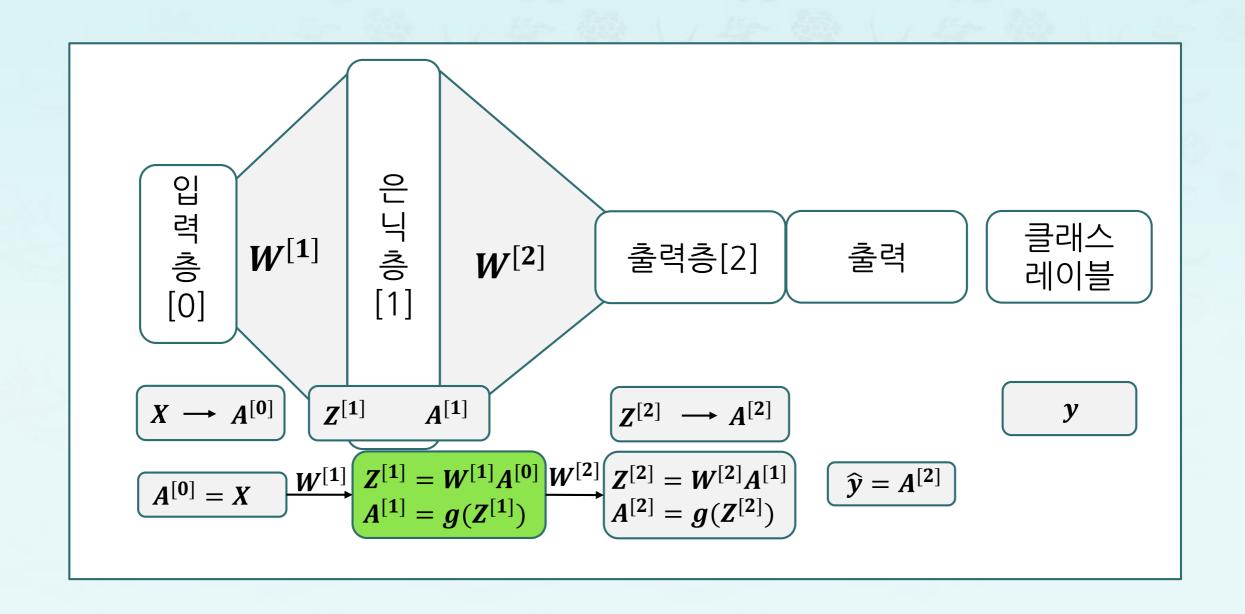
다층 인공신경망 행렬 모델: 입력과 출력

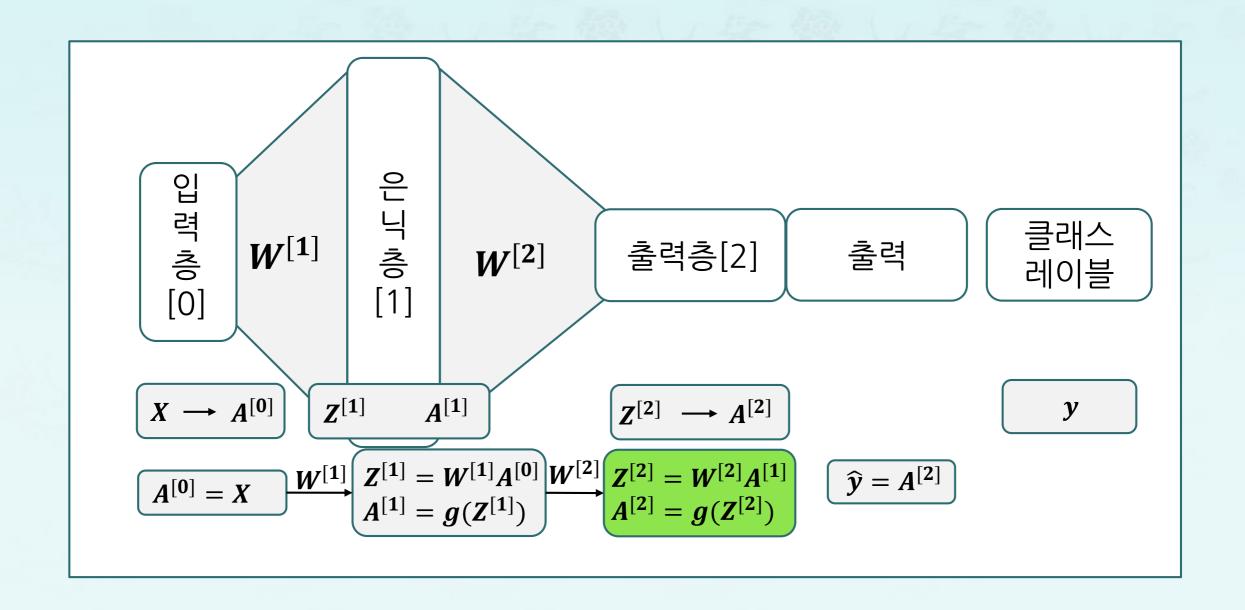


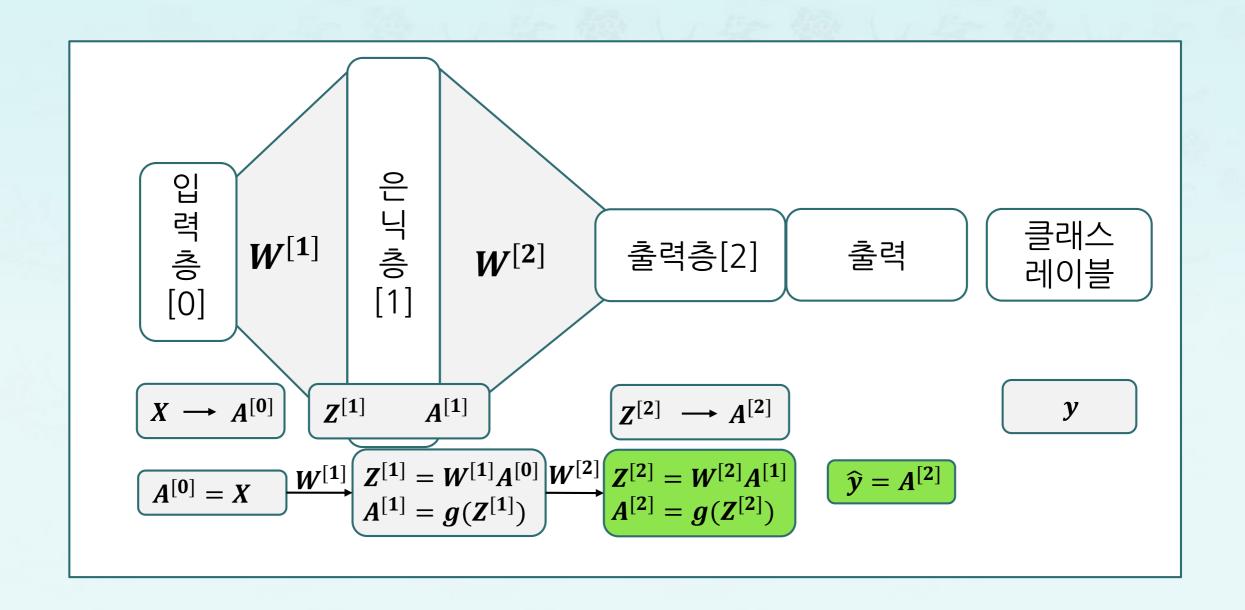
다층 인공신경망 행렬 모델: 입력과 출력

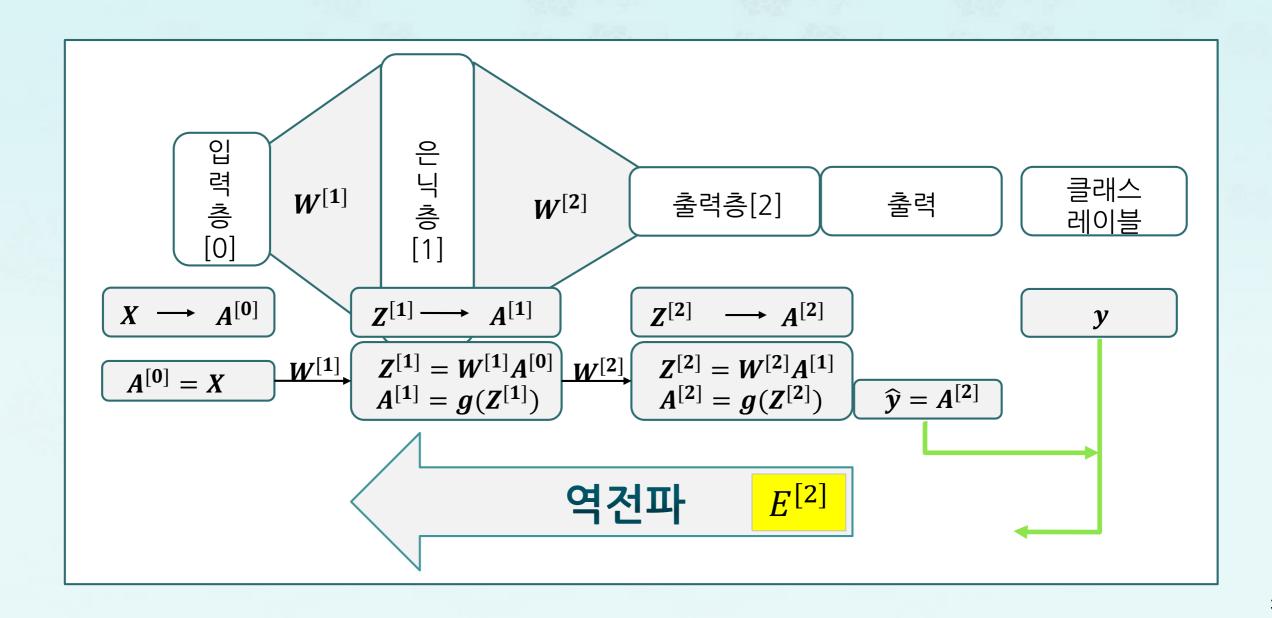


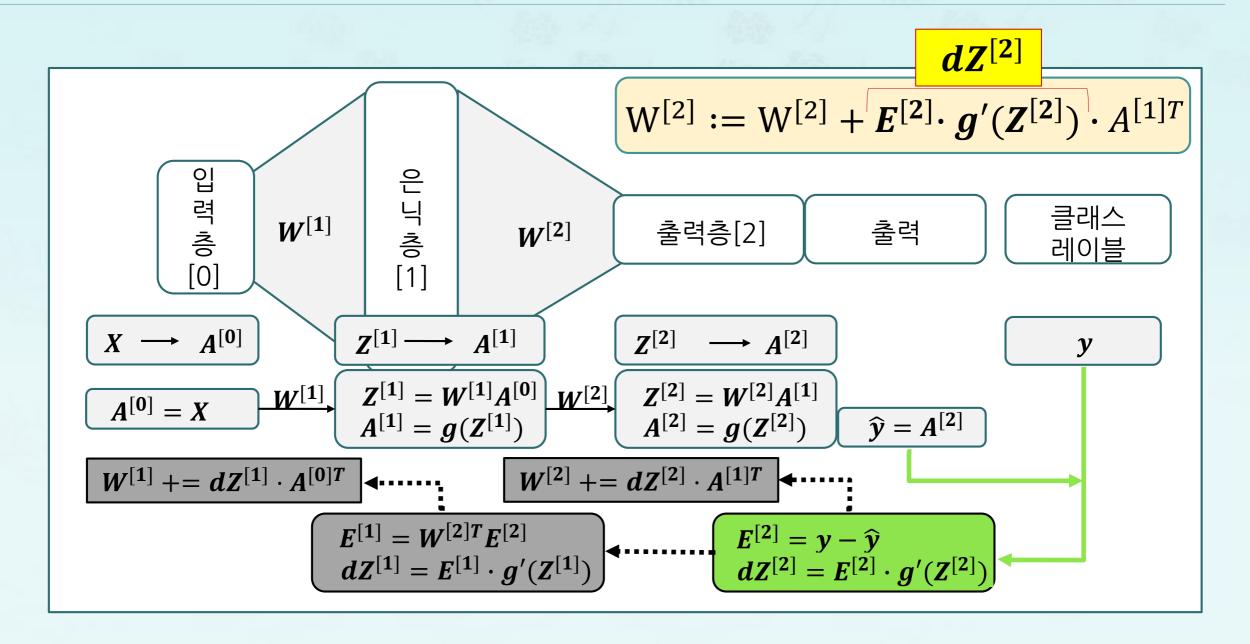


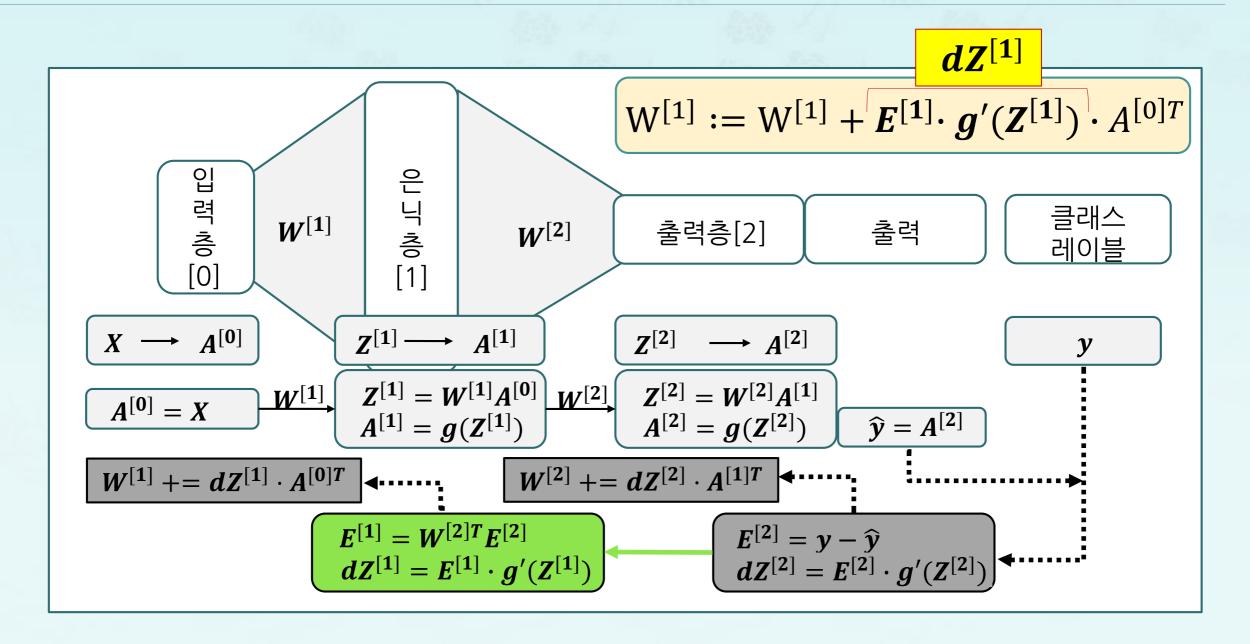




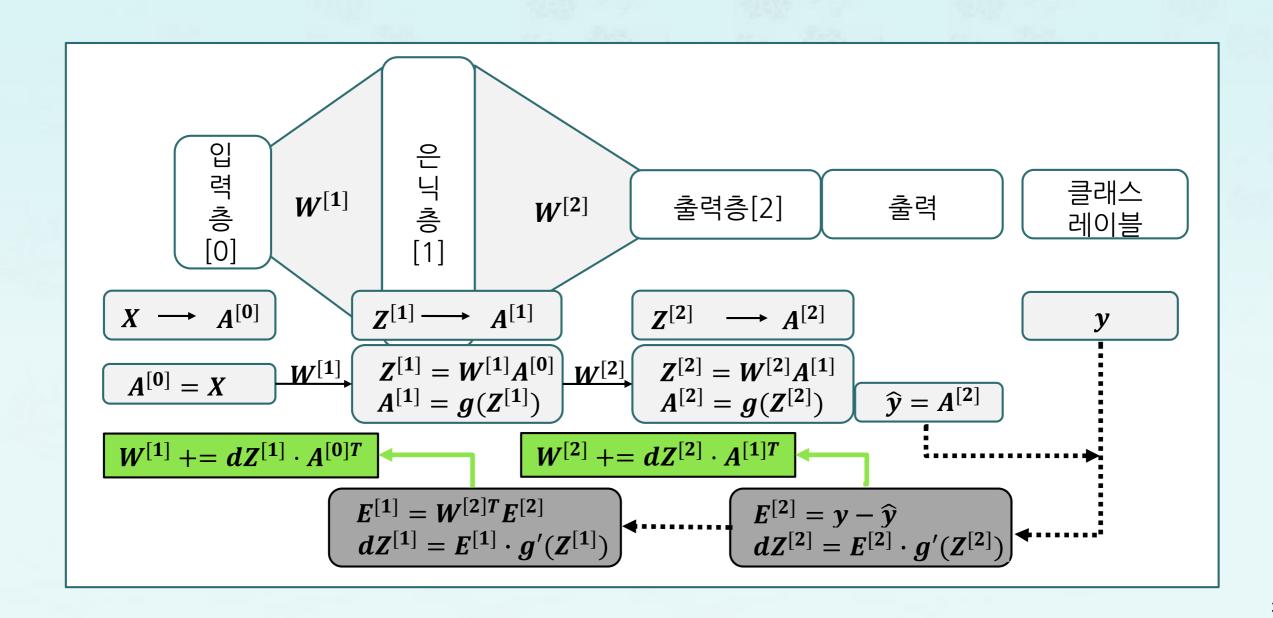








다층 인공신경망 행렬 모델: 역전파



역전파 2: 역전파의 가중치 조정

■ 최종 결과

$$W^{[2]} := W^{[2]} - \alpha \Delta W^{[2]}$$

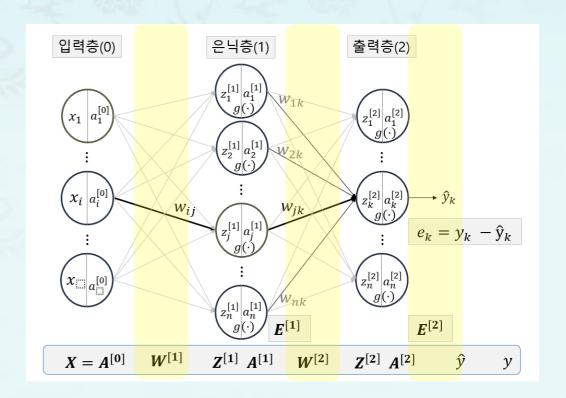
$$= W^{[2]} - \alpha \frac{\partial E}{\partial W^{[2]}}$$

$$= W^{[2]} + \alpha E^{[2]} \cdot g'(Z^{[2]}) \cdot A^{[1]T}$$

$$W^{[1]} := W^{[1]} - \alpha \Delta W^{[1]}$$

$$= W^{[1]} - \alpha \frac{\partial E}{\partial W^{[1]}}$$

$$= W^{[1]} + \alpha E^{[1]} \cdot g'(Z^{[1]}) \cdot A^{[0]T}$$



다층 신경망의 행렬 모델링

- 학습 정리
 - 미분에서 연쇄법칙이 무엇인지 이해하기
 - 오차함수의 행렬로 미분하기
 - 다층 인공 신경망의 행렬 모델을 학습하기

- 차시 예고
 - 10-2 로지스틱 회귀

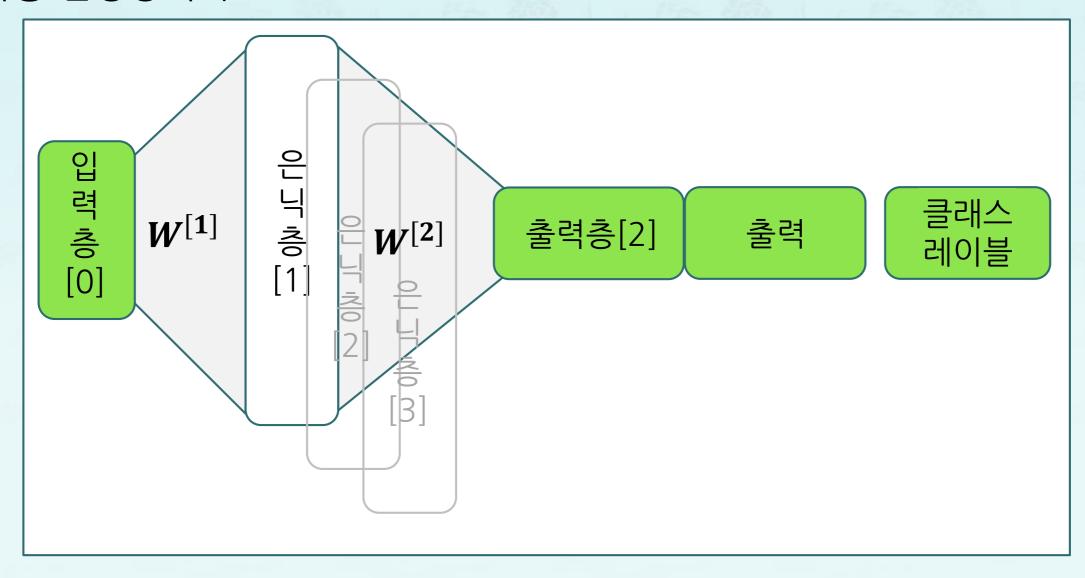
11주차(1/3)

다층 신경망의 행렬 모델링

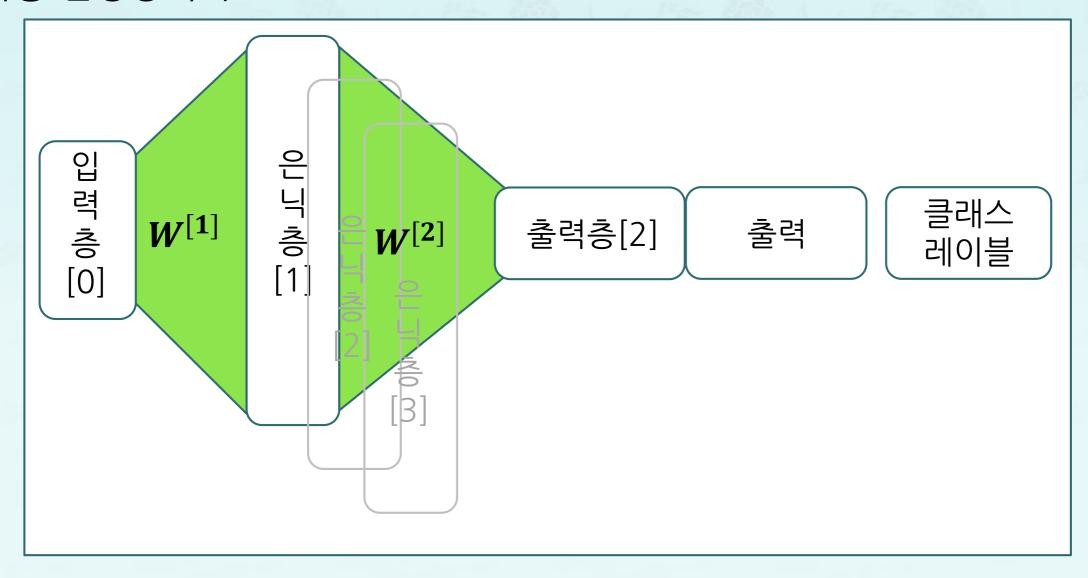
파이썬으로배우는기계학습

한동대학교 김영섭교수

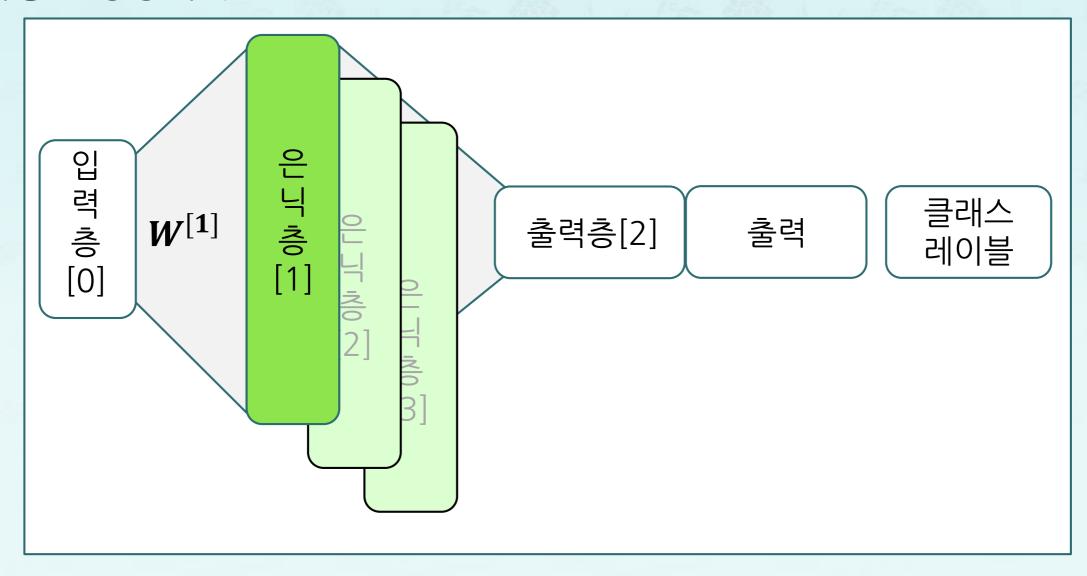
■ 다층 신경망의 구조

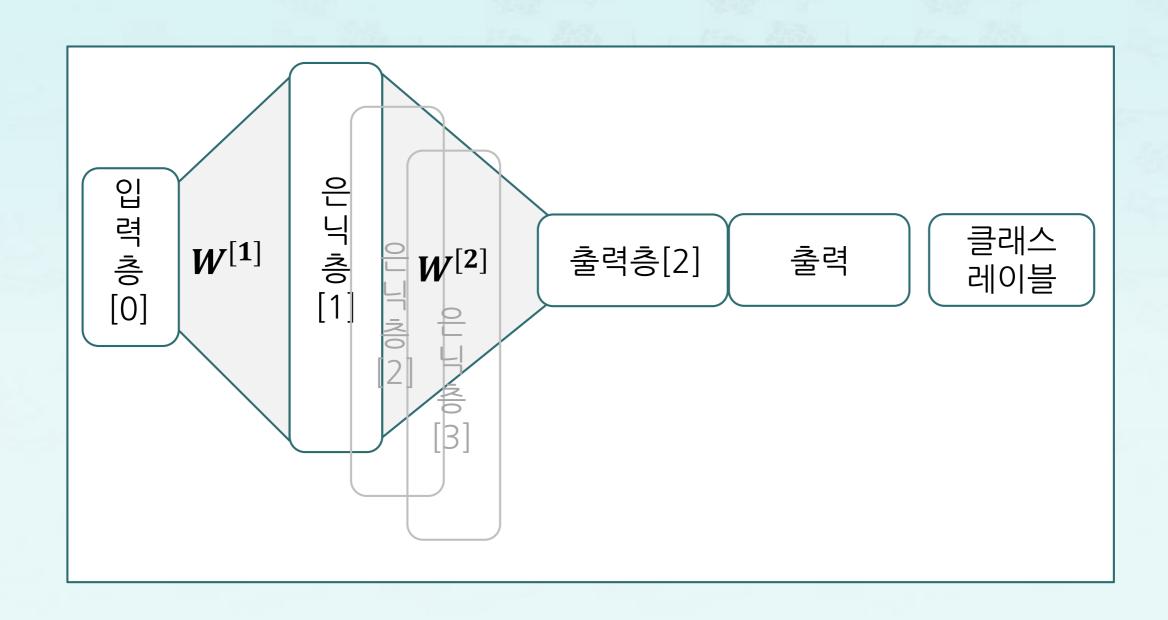


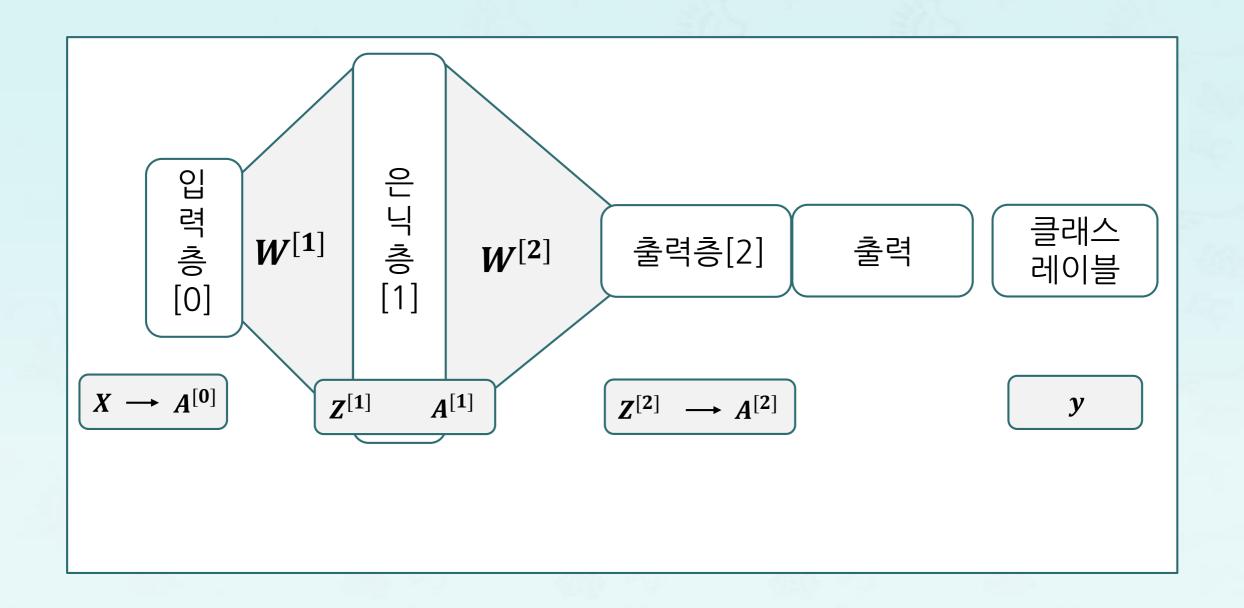
■ 다층 신경망의 구조



■ 다층 신경망의 구조







■ 역전파 모델링

