

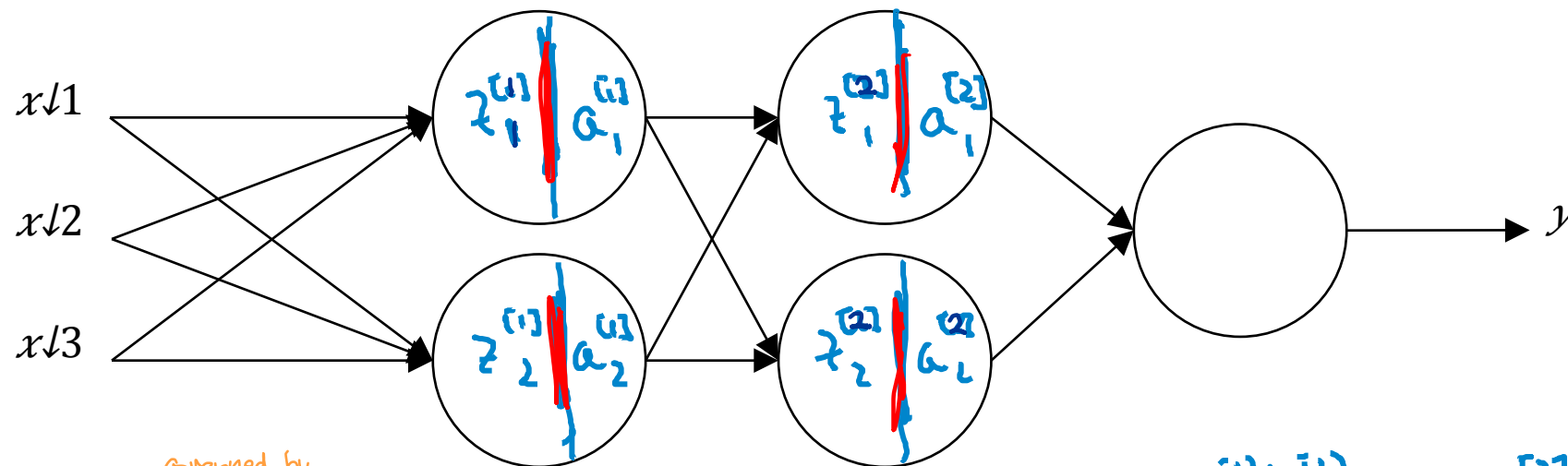


deeplearning.ai

Batch Normalization

Fitting Batch Norm
into a neural network

Adding Batch Norm to a network



governed by

$$X \xrightarrow{W^{(1)}, b^{(1)}} z^{(1)} \xrightarrow{\beta^{(1)}, \gamma^{(1)}} \tilde{z}^{(1)} \xrightarrow{W^{(2)}, b^{(2)}} z^{(2)} \xrightarrow{\beta^{(2)}, \gamma^{(2)}} \tilde{z}^{(2)} \xrightarrow{W^{(3)}, b^{(3)}} z^{(3)} \xrightarrow{\beta^{(3)}, \gamma^{(3)}} \tilde{z}^{(3)} \rightarrow a^{(3)}$$

Batch Norm (BN)

Parameters:

$$\left\{ W^{(1)}, b^{(1)}, W^{(2)}, b^{(2)}, \dots, W^{(L)}, b^{(L)} \right\}$$

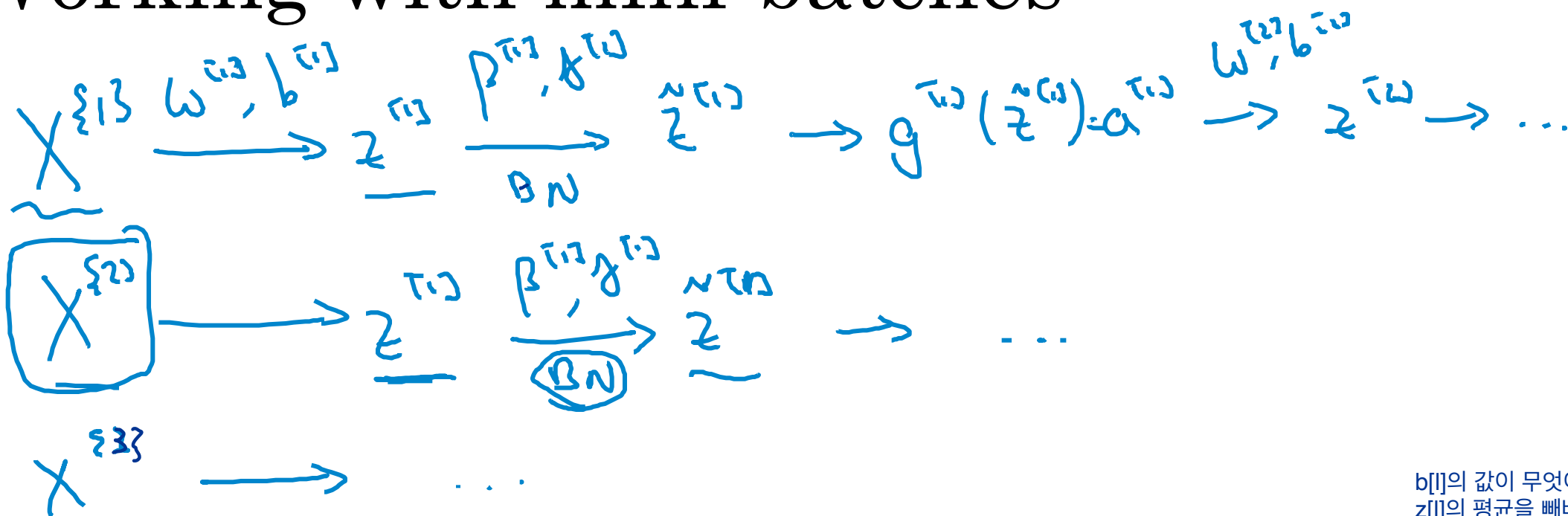
$$\rightarrow \beta^{(1)}, \gamma^{(1)}, \beta^{(2)}, \gamma^{(2)}, \dots, \beta^{(L)}, \gamma^{(L)}$$

$$\rightarrow \beta$$

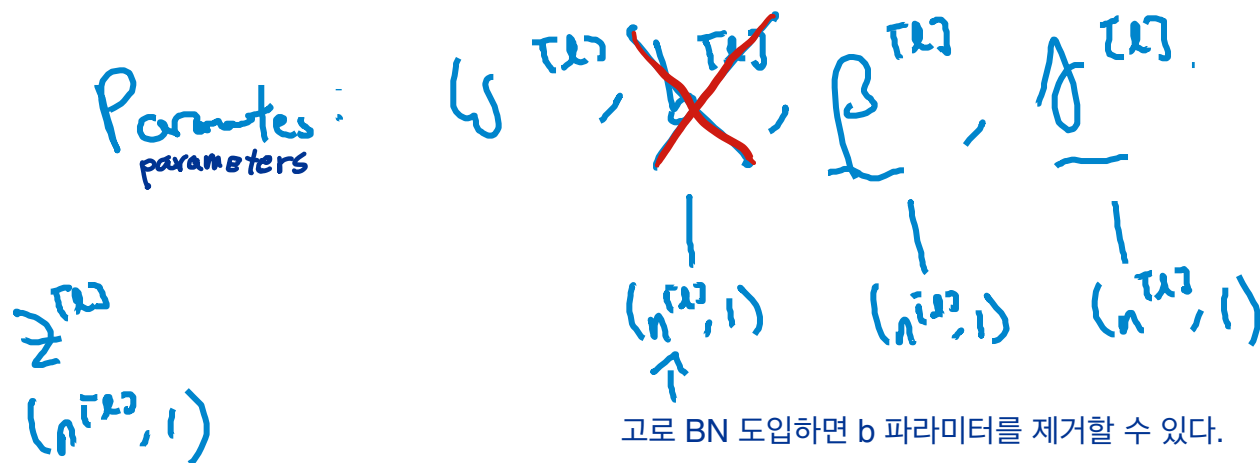
$$d\beta^{(2)} \quad \beta = \beta - \alpha d\beta^{(2)}$$

tf.nn.batch-normalization ←

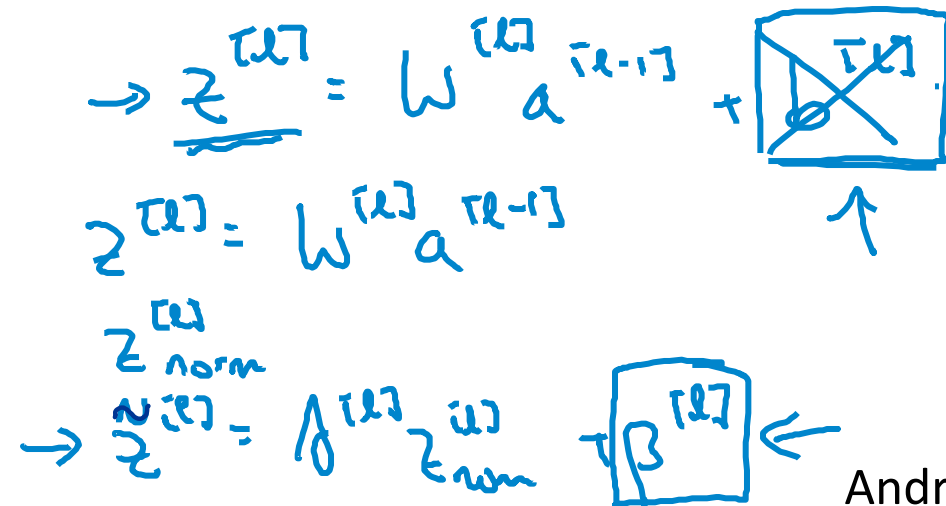
Working with mini-batches



$b[i]$ 의 값이 무엇이건 간에 BN 과정에서 $z[i]$ 의 평균을 빼버리기 때문에 소용없는 값이 되어 버린다.



고로 BN 도입하면 b 파라미터를 제거할 수 있다.



Implementing gradient descent

for $t = 1 \dots \text{num Mini Batches}$

Compute forward pass ^{propagation} on $X^{\{t\}}$.

In each hidden layer, use BN to ^{replace}

Use backprop ^{to} compute $\underline{dw}^{(i)}$, ~~$\underline{dx}^{(i)}$~~ , $\underline{dp}^{(i)}$, $\underline{d\delta}^{(i)}$

Update params
$$\left. \begin{aligned} W^{(i)} &:= W^{(i)} - \alpha dw^{(i)} \\ \beta^{(i)} &:= \beta^{(i)} - \alpha dp^{(i)} \\ \gamma^{(i)} &:= \dots \end{aligned} \right\} \leftarrow$$

$\underline{z}^{(i)}$ with $\tilde{z}^{(i)}$.

Works w/ momentum, RMSprop, Adam.