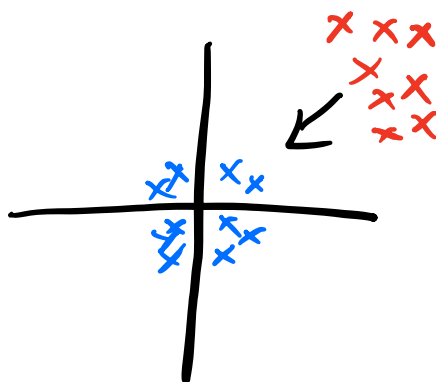
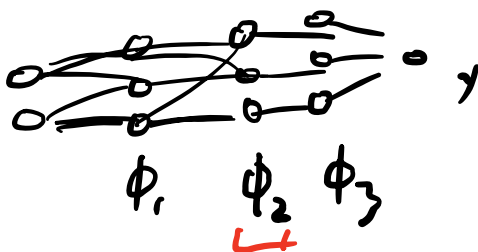


Last time

Normalization: keep data at consistent mean and variance

Advantages $BN(x) = \frac{x - E[x]}{\sqrt{Var[x]}}$

$$\left| \frac{dL}{dw_i} \right| = |x| |\sigma'(xw_i)| \prod_{j=1}^L \left| \frac{d\phi_j}{d\phi_{j-1}} \right|$$

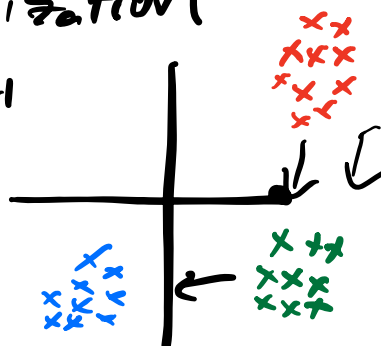


$$\left| \frac{dL}{dw_i} \right| = |\phi_{i-1}| |\sigma'(\phi_{i-1}w_i)| \prod_{j=i+1}^L \left| \frac{d\phi_j}{d\phi_{j-1}} \right|$$

$$BN(\phi_{i-1}) = \frac{\phi_{i-1} - E[\phi_{i-1}]}{\sqrt{Var(\phi_{i-1}) - \epsilon}}$$

w/o normalization

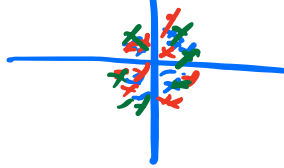
ϕ_{i-1}



$$\phi_{i-1} = \sigma(xw_i)$$

\uparrow
change

W/ Normalization



Batch Norm

$$\begin{aligned} & \text{X} \quad \text{O} \quad E[\phi_{i,1}] = 0 \quad \text{Var} = 1 \\ & \text{---} \quad \text{O} \quad E[\] = 0 \quad \text{Var} = 1 \\ & \text{---} \quad \text{O} \quad \vdots \\ & \text{---} \quad \text{O} \quad \vdots \\ & \text{---} \quad \text{O} \quad \vdots \end{aligned}$$

Layer Norm

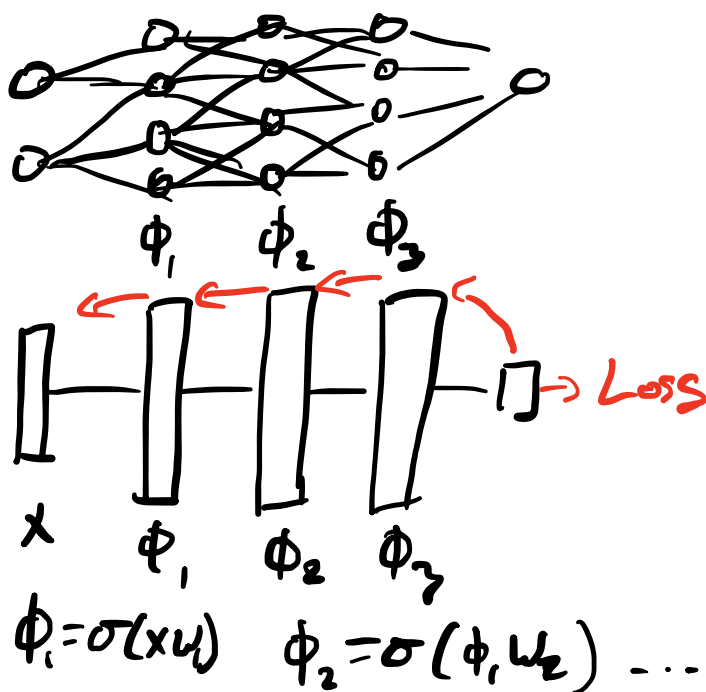
$$\begin{aligned} & E[x] = 0 \\ & \text{Var} \geq 1 \end{aligned}$$

$$N \times d$$

obj # neurons

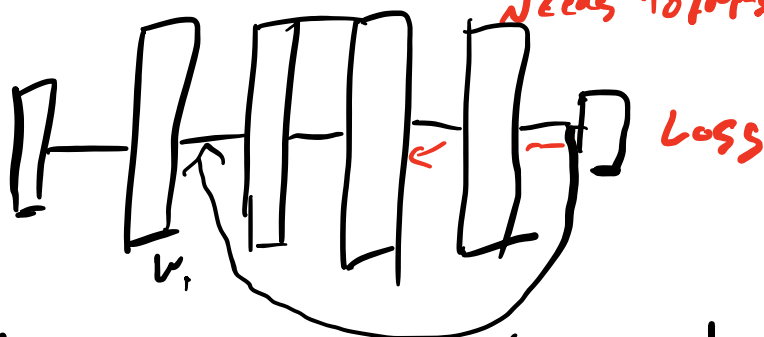
Batch norm

Residual Networks



$$\frac{dL}{dw_1} = x \sigma'(x^T w_1) \frac{d\phi_2}{d\phi_1} \cdot \frac{d\phi_3}{d\phi_2} \dots \frac{dL}{d\phi_L}$$

Needs to pass info about Loss



$$\frac{dL}{dw_1} = x \sigma'(xw_1) \left[\left(\prod_{i=1}^L \frac{d\phi_i}{d\phi_{i-1}} \right) \frac{dL}{d\phi_L} + \frac{dL}{d\phi_1} \frac{d\phi_1}{dw_1} \right]$$

Approach: Residual function

$$\hat{f}(x) \rightarrow f(x) = \hat{f}(x) + x$$

Residual Layer

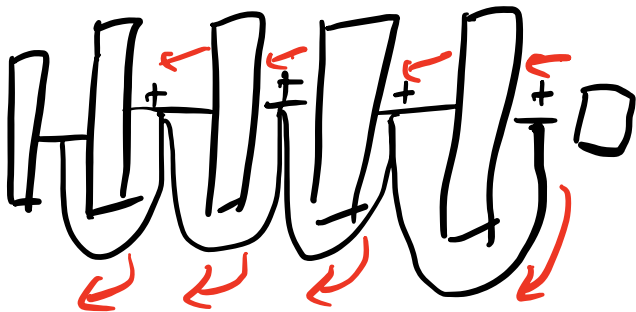
$$\text{Layer } \phi_i = \sigma(\phi_{i-1}^T w_i)$$

Computes next value

$$\text{Residual Layer } \phi_i = \underbrace{\sigma(\phi_{i-1}^T w_i)}_{\hat{\phi}_i} + \underbrace{\phi_{i-1}}_{\frac{d}{d\phi_{i-1}} = 1}$$

$$\hat{\phi}_i = \sigma(\phi_{i-1}^T w_i) = \phi_i - \phi_{i-1}$$

residual



Backprop step

$$\frac{dL}{d\phi_{i-1}} = \frac{dL}{d\phi_i} \cdot \frac{d\phi_i}{d\phi_{i-1}} = \frac{dL}{d\phi_i} \left[\frac{d}{d\phi_{i-1}} \sigma(\phi_{i-1}^T w_i) \right] + \frac{dL}{d\phi_i}$$

$$= \frac{dL}{d\phi_i} \left[\frac{d}{d\phi_{i-1}} (\sigma(\phi_{i-1}^T w_i) + \phi_{i-1}) \right]$$

$$= \frac{dL}{d\phi_i} [w_i \sigma'(\phi_{i-1}) + 1]$$

w/o residual

$$\frac{dL}{dw_i} = x \sigma'(xw_i) \prod_{i=1}^L \left(\frac{d\phi_i}{d\phi_{i-1}} \right) \frac{dL}{d\phi_L}$$

w/ residual

$$\frac{dL}{dw_1} = x \sigma'(xw_1) \prod_{i=1}^L \left(\frac{d\hat{\phi}_i}{d\phi_{i-1}} + 1 \right) \frac{dL}{d\phi_L}$$