

Forward

$h^l = f_l(h^{l-1})$ and $y = l \circ f_d \circ \dots \circ f_1(x)$

Gradient Problem

Gradient Exploding

- 1. If w is initialized with > 1 ,
- 2. get too many layers

Weights get too big for float16

Parameter Initialization

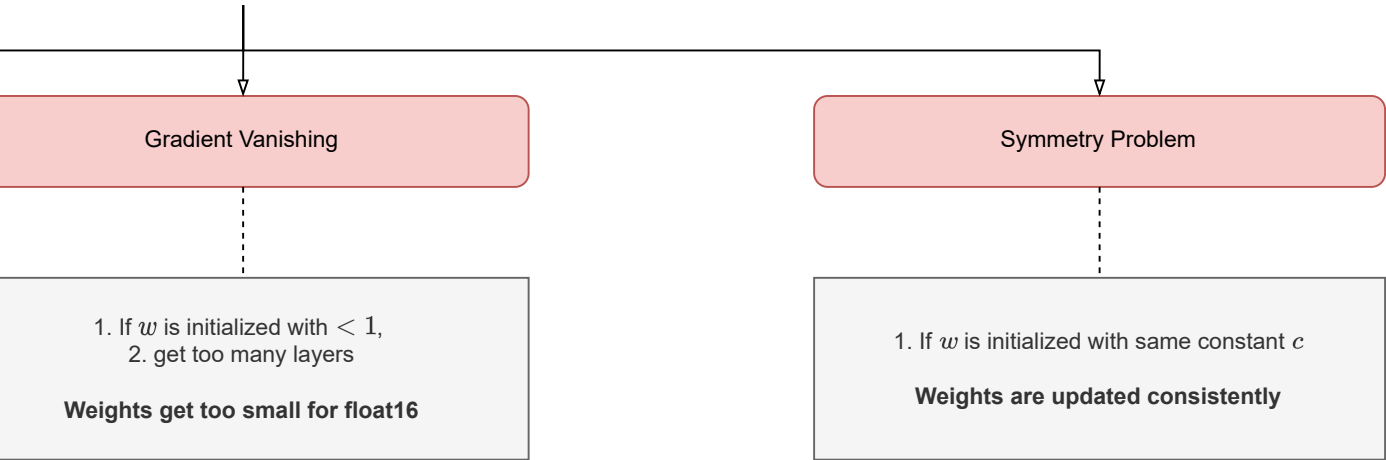
Target:

$$\mathbb{E}\left[\frac{\partial l}{\partial h^t}\right] = 0$$
$$Var\left[\frac{\partial l}{\partial h^t}\right] = b$$

Gradient Problem

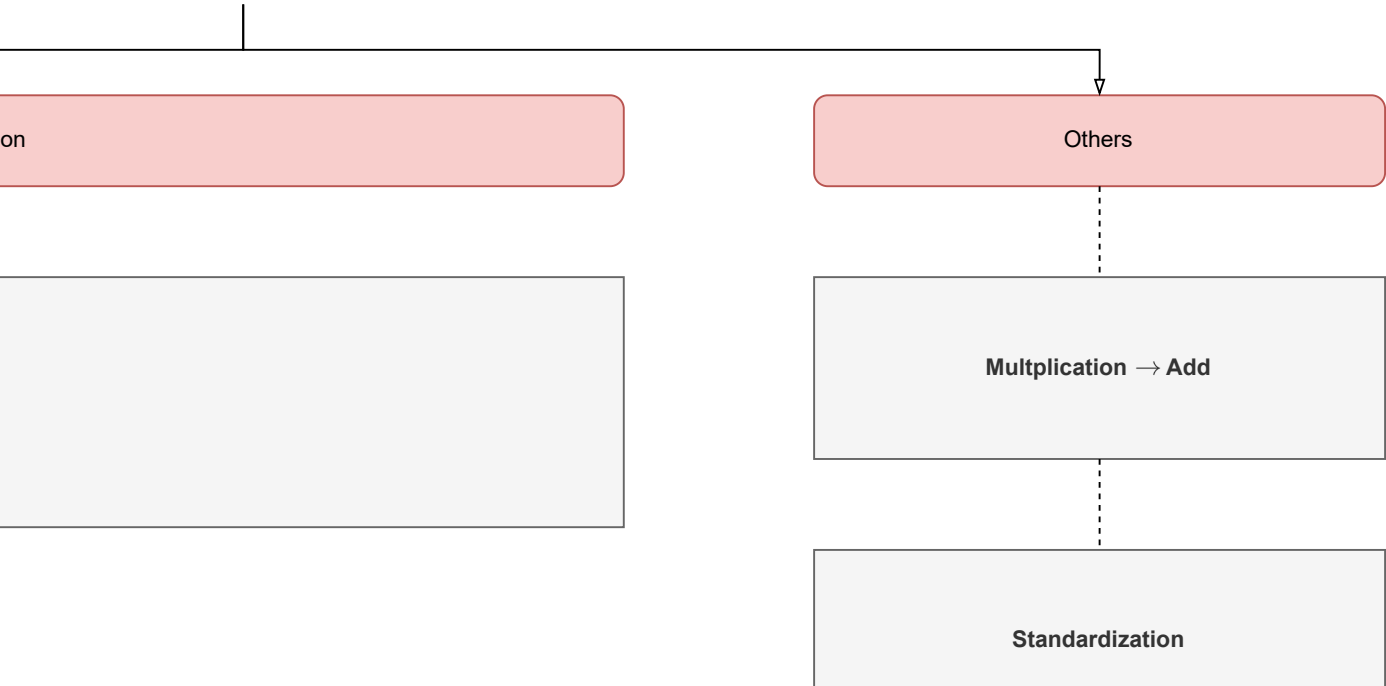
Backward

$$\frac{\partial l}{\partial W^t} = \frac{\partial l}{\partial h^d} \frac{\partial h^d}{\partial h^{d-1}} \cdots \frac{\partial h^{t+1}}{\partial h^t} \frac{\partial h^t}{\partial W^t}$$
$$= M^L \dots M^{l+1} v^l$$

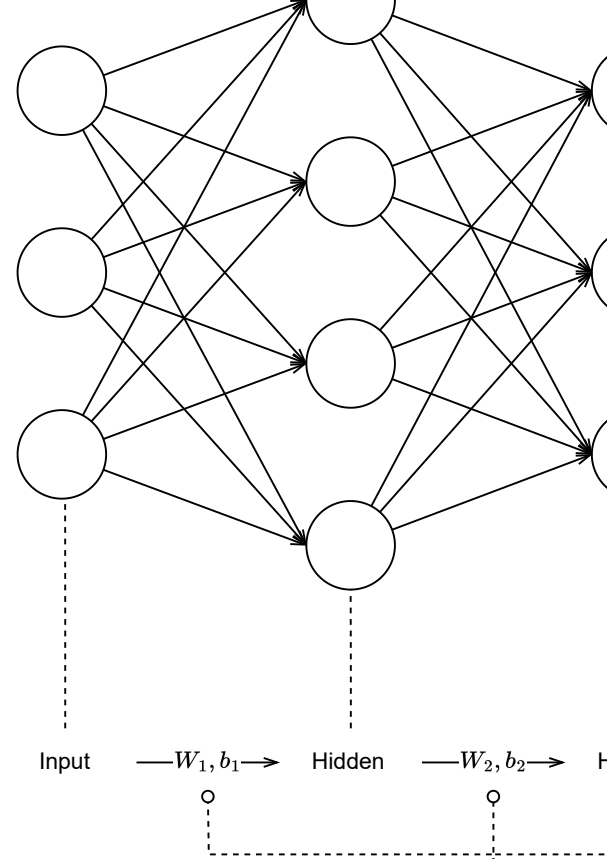


Target

Let gradients lie in proper interval



Solution



MLP example (h = hidden layer weights)

Condition:

1. $w_{i,j}^t$ is of i.i.d, **we get:** $\mathbb{E}\left[\frac{\partial l}{\partial h^t}\right] = 0$ $Var\left[\frac{\partial l}{\partial h^t}\right] = b$
2. h_i^{t-1} is independent to $w_{i,j}^t$
3. Dont consider activation function

Expectation and Variance of every layers' weights:

1. Forward $\rightarrow \mathbb{E}[h_i^t] = 0$, $Var[h_i^t] = n_{t-1}\gamma_t Var[h_i^{t-1}] \rightarrow n_{t-1}\gamma_t =$
2. Backward $\rightarrow \mathbb{E}\left[\frac{\partial l}{\partial h_i^{t-1}}\right] = 0$, $Var\left[\frac{\partial l}{\partial h_i^{t-1}}\right] = n_t\gamma_t Var\left[\frac{\partial l}{\partial h_i^t}\right] \rightarrow$

Hard to fullfill two of them

Solution: **Xavier Initialization**

1. **Let** $\gamma_t \frac{n_{t-1} + n_t}{2} = 1 \rightarrow \gamma_t = \frac{2}{n_{t-1} + n_t}$
2. **We get** $\mathcal{N}\left(0, \sqrt{\frac{2}{n_{t-1} + n_t}}\right)$, $\mathcal{U}\left(-\sqrt{\frac{6}{n_{t-1} + n_t}}, \sqrt{\frac{6}{n_{t-1} + n_t}}\right)$

What if to consider activation function?

