

Lecture 11

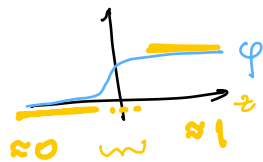
Q1 done ✓
Hw5 ^{out} Wed 2/25
LIV - T.B.
extended

1. How does a nn represent a function f ? ✓

2. Backprop. analysis ←

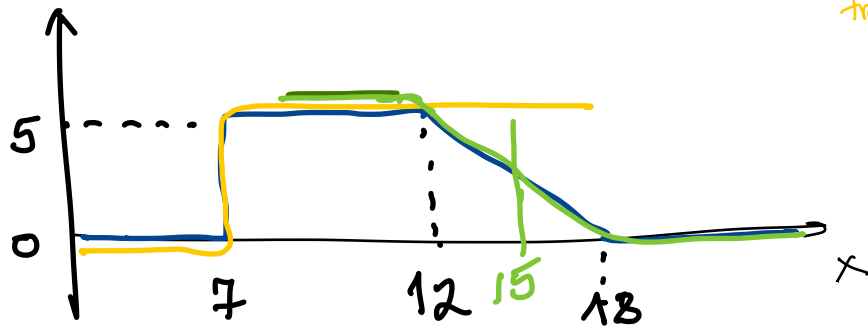
$$f = \varphi(w^T x)$$

$$x^{(t)} = \varphi(w_i^{(t)} x^{(t-1)})$$



$z = w^T x \Rightarrow$ steeper slope for $\|w\|$ large $\Rightarrow$

"transition" smaller



$$w_1 = \text{large}_1 > 0$$

$$w_0 = -7 \cdot 1000$$

$$z_1 = 1000(x - 7)$$

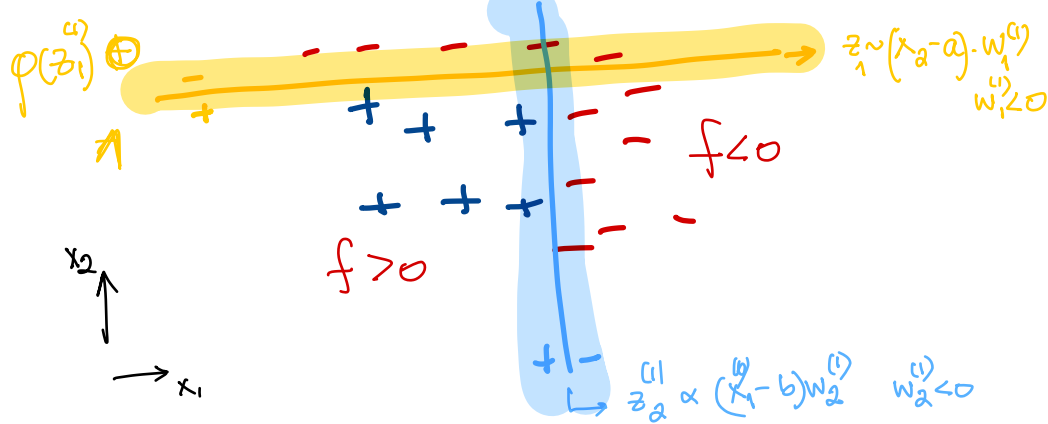
$$w_1^2 = 1$$

$$w_0^2 = -15$$

$$z_2 = x - 15$$

$$f = \frac{w_1^3}{w_2^3} \cdot \varphi(z_1)$$

$$-5 \varphi(z_2)$$



$\leftarrow \dots$
 -10000

$$f = w_1^{(2)} \phi(z_1) + w_2^{(2)} \phi(z_2) - \underbrace{1.5}_{\approx 1} \phi(z_3)$$

$z_1^{(2)} \leftarrow f$

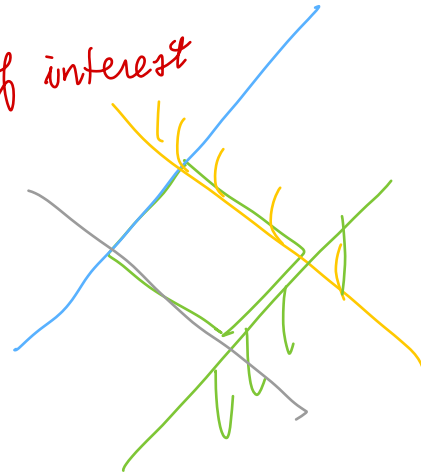
$z_{1:m_2}^{(2)}$

$\phi(z_2)$

0

$z_3 = x_1 + 10000 \Rightarrow \phi(z_3) \approx 1$ on the region of interest

$$w_1^{(1)} = w_2^{(1)} = 1$$



Analysis of Backprop

$$W = \{W^{(l)} \in \mathbb{R}^{m_l \times m_{l-1}}, l=1:L\} \quad O(p \cdot n) \text{ op/iter}$$

I Computation / iteration ?

T How many steps to convergence

S Saturation ($\phi'(z^{(l)}) \approx 0$)

L Local optima — (minima)

O Overfitting

$T \sim ?$
T.p.n for training

$$\left[\sum_{l=1}^L m_l m_{l-1} \text{ parameters} \right] = P$$

1) $L=2 \quad m_1=1000, m_2=1$

$$m_0 = d = 1M$$

$$\# \text{params} = 10^9 + \dots$$

2) $m_0 = m_1 = \dots = m_{L-1} = 100$

$$L=100$$

$$\# \text{params} \approx 100^2 \cdot 100 \cdot 10^6$$

"T" ← GD "slow" compared to 2nd order
↙ saturation

Optimization "hierarchy"

Newton

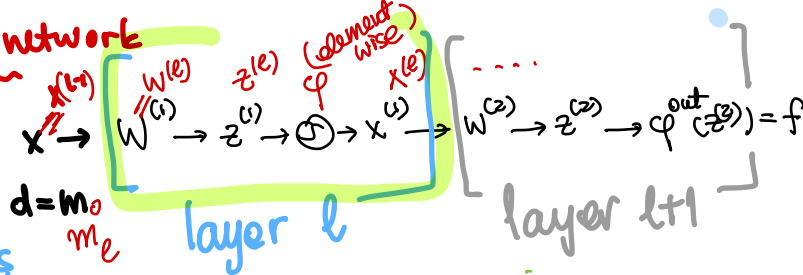
2nd order

Grad

1st order

$$\|W^t - W^\infty\| \propto \beta^t \quad \beta < 1$$

L layer network



Parameters

$$W = \{ W^{(l)} \in \mathbb{R}^{m_l \times m_{l-1}}, l = 1:L \}$$

Chain Rule:

$$\frac{\partial f}{\partial z^{(2)}} = \phi'_{\text{out}}(z^{(2)})$$

$$\frac{\partial z^{(2)}}{\partial w^{(2)}} = x^{(1)}$$

$$z = W^T x$$

$$\frac{\partial z^{(2)}}{\partial x^{(1)}} = w^{(2)}$$

$$\frac{\partial \mathcal{L}}{\partial w^{(2)}} = \frac{\partial \mathcal{L}}{\partial f} \cdot \frac{\partial f}{\partial z^{(2)}} \cdot \frac{\partial z^{(2)}}{\partial w^{(2)}} = -(y_{\text{target}} - \phi_{\text{out}}) \cdot \phi'_{\text{out}}(z^{(2)}) \cdot x^{(1)}$$

$$\frac{\partial x_k^{(1)}}{\partial z_k^{(1)}} = \phi'(z_k^{(1)})$$

$$\frac{\partial x_k^{(1)}}{\partial z_k^{(1)}} = \phi'(z_k^{(1)}) \in \mathbb{R}^m$$

$$z_k^{(1)} = w_k^T x^{(1)} \quad \text{row } k \text{ of } W^{(1)}$$

$$\frac{\partial z_k^{(1)}}{\partial x_k^{(1)}} = w_k^{(1)}$$

$$\frac{\partial z_k^{(1)}}{\partial w_k^{(1)}} = x_k^{(1)}$$

$$\Rightarrow \frac{\partial \mathcal{L}}{\partial w_k^{(1)}} = \frac{\partial \mathcal{L}}{\partial f} \cdot \frac{\partial f}{\partial x_k^{(1)}} \cdot \frac{\partial x_k^{(1)}}{\partial z_k^{(1)}} \cdot \frac{\partial z_k^{(1)}}{\partial w_k^{(1)}} = -(y_{\text{target}} - \phi_{\text{out}}) \cdot \phi'_{\text{out}}(z^{(2)}) \cdot \phi'(z_k^{(1)}) \cdot x_k^{(1)}$$

NEW

$$\frac{\partial f}{\partial x^{(l+1)}}$$

$\times n$ data points (x_i, y_i)

$$\frac{\partial f}{\partial x^{(l)}} = \frac{\partial f}{\partial x^{(l+1)}} \cdot \frac{\partial x^{(l+1)}}{\partial x^{(l)}}$$

$$\frac{\partial x_j^{(l+1)}}{\partial x_i^{(l)}} = \phi'(z_j^{(l+1)}) \cdot \frac{\partial z_j^{(l+1)}}{\partial x_i^{(l)}} = w_{ji}^{(l+1)}$$

$$z_j^{(l+1)} = w_j^{(l+1)T} \cdot x^{(l)}$$

$$\frac{\partial f}{\partial x^{(l)}} = \frac{\partial f}{\partial z^{(l+1)}} \cdot \frac{\partial z_j^{(l+1)}}{\partial x^{(l)}} = \frac{\partial f}{\partial z^{(l+1)}} \cdot w_j^{(l+1)}$$