

## 1 Derivatives of simple functions

Compute the derivatives of the following simple functions used as non-linearities in neural networks.

1.  $\sigma(x) = \frac{1}{1+e^{-x}}$
2.  $\text{ReLU}(x) = \max(x, 0)$
3.  $\tanh(x) = \frac{\sinh(x)}{\cosh(x)} = \frac{e^x - e^{-x}}{e^x + e^{-x}}$

## 2 Backpropagation Practice (self-study)

1. Chain rule of multiple variables: Assume that you have a function given by  $f(x_1, x_2, \dots, x_n)$ , and that  $g_i(w) = x_i$  for a scalar variable  $w$ . How would you compute  $\frac{d}{dw} f(g_1(w), g_2(w), \dots, g_n(w))$ ? What is its computation graph?(also sometimes referred to as a pictorial representation of the network). This is a directed graph of decomposed function computations, with the function at one end, and the variables  $W, b, x, y$  at the other end, where  $b = [b_1, \dots, b_n]$
2. Let  $w_1, w_2, \dots, w_n \in \mathbb{R}^d$ , and we refer to these variables together as  $W \in \mathbb{R}^{n \times d}$ . We also have  $x \in \mathbb{R}^d$  and  $y \in \mathbb{R}$ . Consider the function

$$f(W, x, y) = \left( y - \sum_{i=1}^n \phi(w_i^\top x + b_i) \right)^2.$$

Write out the function computation graph.

3. Suppose  $\phi(x) = \sigma(x)$  (from problem 1a). Compute the partial derivatives  $\frac{\partial f}{\partial w_i}$  and  $\frac{\partial f}{\partial b_i}$ . Use the computational graph you drew in the previous part to guide you.
4. Write down a single gradient descent update for  $w_i^{(t+1)}$  and  $b_i^{(t+1)}$ , assuming step size  $\eta$ . Your answer should be in terms of  $w_i^{(t)}, b_i^{(t)}, x$ , and  $y$ .
5. (optional) Define the cost function

$$\ell(x) = \frac{1}{2} \|W^{(2)} \Phi(W^{(1)} x + b) - y\|_2^2, \tag{1}$$

where  $W^{(1)} \in \mathbb{R}^{d \times d}$ ,  $W^{(2)} \in \mathbb{R}^{d \times d}$ , and  $\Phi : \mathbb{R}^d \rightarrow \mathbb{R}^d$  is some nonlinear transformation. Compute the partial derivatives  $\frac{\partial \ell}{\partial x}$ ,  $\frac{\partial \ell}{\partial W^{(1)}}$ ,  $\frac{\partial \ell}{\partial W^{(2)}}$ , and  $\frac{\partial \ell}{\partial b}$ .

6. (optional) Suppose  $\Phi$  is the identity map. Write down a single gradient descent update for  $W_{t+1}^{(1)}$  and  $W_{t+1}^{(2)}$  assuming step size  $\eta$ . Your answer should be in terms of  $W_t^{(1)}$ ,  $W_t^{(2)}$ ,  $b_t$  and  $x, y$ .