

# CS 4501: Optimization - Assignment 7

Your name and email

Consider an  $n$ -layer feedforward neural network (FNN) parameterized by

$$W = (W_0, W_1, \dots, W_{n-1}),$$
$$b = (b_0, b_1, \dots, b_{n-1}),$$

where,

$$W_i \in \mathbb{R}^{d_i \times d_{i+1}},$$
$$b_i \in \mathbb{R}^{d_{i+1}}, \quad i = 0, 1, \dots, n-1.$$

Here the sequence  $d \doteq (d_0, d_1, \dots, d_n)$  specifies the number of units in each layer, with  $d_0$  being the size of the input  $x$  and  $d_n$  being the size of the output. Let  $\sigma : \mathbb{R} \rightarrow \mathbb{R}$  be the activation function. Notably, we use the convention that if  $x$  is matrix / vector,  $\sigma(x)$  denotes a matrix / vector obtained by elementwise application of  $\sigma$  in  $x$ .

Given an input  $x \in \mathbb{R}^{d_0}$ , the output  $f(x; W, b) \in \mathbb{R}^{d_n}$  is given by

$$x_0 \doteq x$$
$$z_{k+1} \doteq W_k^\top x_k + b_k,$$
$$x_{k+1} \doteq \sigma(z_{k+1}), \quad k = 0, 1, \dots, n-1,$$
$$f(x; W, b) \doteq x_n.$$

Suppose the target output is  $y \in \mathbb{R}^{d_n}$ , the loss function is then defined as

$$L(x, y) = \frac{1}{2} \|f(x; W, b) - y\|_2^2.$$

## Task 1 (5pt): Compute the gradients analytically

*Proof.* (SZ: You can use recursive expression in expressing the gradients.)

$$\frac{dL(x, y)}{dW_k} =$$
$$\frac{dL(x, y)}{db_k} =$$

□

## Task 2 (25pt): Compute the gradients in python

(SZ: You cannot use any auto-diff software in your submission. You can of course use auto-diff software to verify your computation. Just remember to delete them in your submission.)