

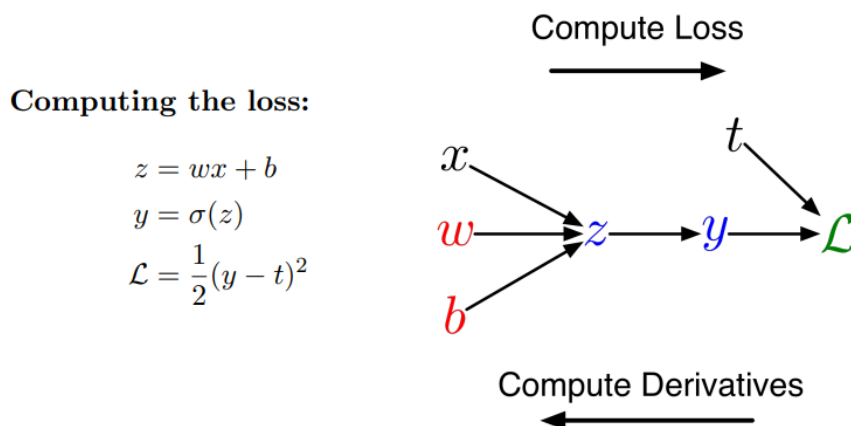
# CSCI 416/516 Homework #3

DUE: November 27, 2024, at 11:59 pm

**CSCI 416/516:** Each Problem begins with an allocation of points, represented as [ $u$  pts/ $g$  pts]. If you are registered in CSCI 416, you can receive up to  $u$  pts on this Problem; if you are registered in CSCI 516, you can receive up to  $g$  pts on this Problem. The last Problem is optional for undergraduates (CSCI 416) but required for graduates (CSCI 516). **Write down which session you are in / are you a graduate or undergraduate student.**

**Submission:** You need to submit both your homework report (answers to the Problems) as a PDF file through Blackboard. Please show the work on how you reach the conclusion for each question except the multiple choice questions and Yes/No (True/False) questions.

- **Problem 1 [2 pt(s)/1 pt(s)]: Backpropagation.** What does the back pass look like (in terms of the error signals of parameters/activations, such as but not limited to  $b, w$ ), given the illustrated forward pass in Figure below? Show your work on how the conclusion is reached.



- **Problem 2 [4 pt(s)/2 pt(s)]: Activation Function.** Suppose we have a neural network with one hidden layer as represented in Equation 1 whereas  $g(z)$  is defined as  $\mathbb{I}(z \geq 0)$ . Which of the following functions can be exactly represented as the activation function  $g$ ? Indicate Yes or No to each choice.

$$f(x) = w_0 + \sum_i w_i h_i(x); h_i(x) = g(b_i + v_i x) \quad (1)$$

- Hard Threshold Activation Function
  - Rectified Linear Unit
  - Identity Activation Function
  - Logistic Activation Function
- **Problem 3 2 pt(s)/2 pt(s): 1D Convolution.**  
 Given a kernel  $k = \{2, 9, 3\}$  and a vector  $v = \{1, 2, 3\}$ , what is the result of  $k * v$ ? After getting the result  $k * v$ , what is the new result after applying ReLU on  $k * v$ ? Show the work that leads to your conclusion.
  - **Problem 4 [2 pt(s)/2 pt(s)]: Attention & Transformers.**  
 What's the relationship between the Scaled Dot-Product Attention and Multi-Head Attention, in the transformer architecture?
  - **Problem 5 [3 pt(s)/3 pt(s)]: MLP.**  
 Which of the following statements about multi-layer perceptrons (MLPs) is incorrect?
    - A - An MLP with one hidden layer and a sufficient number of neurons can approximate any continuous function.
    - B - The activation function used in the hidden layers of an MLP must always be non-linear.
    - C - Dropout is a regularization technique used to prevent overfitting by randomly setting some neuron outputs to zero during training.
    - D - MLPs require all input features to be numerical, and categorical data must be transformed using techniques like one-hot encoding or embeddings.