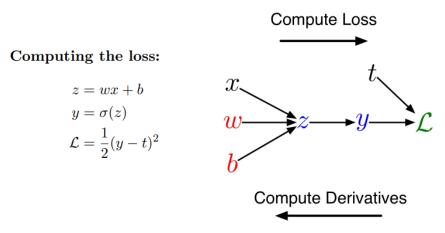
CSCI 416/516 Homework #4

DUE: December 04, 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 2 Problems are 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 [4 pt(s)/2 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 $\mathbb{I}(z \ge 0)$. Which of the following functions can be exactly represented as the activation function g? Indicate Yes or No to each choice.

$$f(x) = w_0 + \sum_i w_i h_i(x); h_i(x) = g(b_i + v_i x)$$
(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 (bonus) [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 (bonus) [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.