

Name: \_\_\_\_\_

## We Recur!

7. (12 points) We have seen in class recurrent neural networks (RNNs) that are structured as:

$$\begin{aligned}z_t^1 &= W^{ss}s_{t-1} + W^{sx}x_t \\s_t &= f_1(z_t^1) \\z_t^2 &= W^o s_t \\p_t &= f_2(z_t^2)\end{aligned}$$

where we have set biases to zero. Here  $x_t$  is the input and  $y_t$  the actual output for  $(x_t, y_t)$  sequences used for training, with  $p_t$  as the RNN output (during or after training).

Assume our first RNN, call it RNN-A, has  $s_t, x_t, p_t$  all being vectors of shape  $2 \times 1$ . In addition, the activation functions are simply  $f_1(z) = z$  and  $f_2(z) = z$ .

- (a) For RNN-A, give dimensions of the weights:

$W^{ss}$ : \_\_\_\_\_       $W^{sx}$ : \_\_\_\_\_       $W^o$ : \_\_\_\_\_

- (b) We have finished training RNN-A, using some overall loss  $J = \sum_t Loss(y_t, p_t)$  given the per-element loss function  $Loss(y_t, p_t)$ . We are now interested in the derivative of the overall loss with respect to  $x_t$ ; for example, we might want to know how sensitive the loss is to a particular input (perhaps to identify an outlier input). What is the derivative of overall loss at time  $t$  with respect to  $x_t$ ,  $\partial J / \partial x_t$ , with dimensions  $2 \times 1$ , in terms of the weights  $W^{ss}, W^{sx}, W^o$  and the input  $x_t$ ? Assume we have  $\partial Loss / \partial z_t^2$ , with dimensions  $2 \times 1$ . Use  $*$  to indicate element-wise multiplication.

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Now consider a modified RNN, call it RNN-B, that does the following:

$$\begin{aligned}z_t^1 &= W^{ssx} \begin{bmatrix} s_{t-1} \\ x_t \end{bmatrix} \\s_t &= z_t^1 \\z_t^2 &= W^{ox} \begin{bmatrix} s_t \\ x_t \end{bmatrix} \\p_t &= f_2(z_t^2)\end{aligned}$$

where  $s_t, x_t, p_t$  are all vectors of shape  $2 \times 1$ ,  $\begin{bmatrix} s_{t-1} \\ x_t \end{bmatrix}$  and  $\begin{bmatrix} s_t \\ x_t \end{bmatrix}$  are vectors of shape  $4 \times 1$ .

(c) For RNN-B, give dimensions of the weights:

$W^{ssx}$ : \_\_\_\_\_       $W^{ox}$ : \_\_\_\_\_

(d) Imagine we are using RNN-B to generate a description sentence given an input word, as in language modeling. The input is a single  $2 \times 1$  vector embedding,  $x_1$ , that encodes the input word. The output will be a sequence of words  $p_1, p_2, \dots, p_n$  that provide a description of that word. In this setting, what would be an appropriate activation function  $f_2$ ?

(e) Continuing with RNN-B for one-to-many description generation using our language modeling approach, we calculate  $p_1$  in a forward pass. How do we calculate  $x_2$  (what is  $x_2$  equal to)?

(f) For RNN-B, we are also interested in the derivative of loss at time  $t$  with respect to  $x_t$ ,  $\partial Loss / \partial x_t$ . Indicate all of the following that are true about RNN-B, and the derivative of loss with respect to  $x_t$ :

- $\partial Loss / \partial x_t$  depends on  $W^{ox}$
- $\partial Loss / \partial x_t$  depends on all elements of  $W^{ox}$
- $\partial Loss / \partial x_t$  depends on  $W^{ssx}$
- $\partial Loss / \partial x_t$  depends on all elements of  $W^{ssx}$