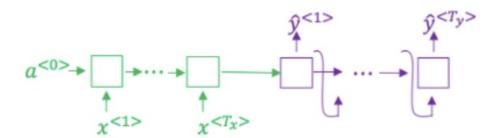
# Sequence models & Attention mechanism

Quiz, 10 questions

1. Consider using this encoder-decoder model for machine translation.



This model is a "conditional language model" in the sense that the encoder portion (shown in green) is modeling the probability of the input sentence x.

|         | True  |  |
|---------|-------|--|
|         | False |  |
| Correct |       |  |

- 2. In beam search, if you increase the beam width B, which of the following would you expect to be true? Check all that apply.
  - Beam search will run more slowly.

Correct

Beam search will use up more memory.

Correct

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| ı |

Beam search will generally find better solutions (i.e. do a better job maximizing  $P(y \mid x)$ )

## Correct



Beam search will converge after fewer steps.

# Un-selected is correct

3. In machine translation, if we carry out beam search without using sentence normalization, the algorithm will tend to output overly short translations.



True

#### Correct



4. Suppose you are building a speech recognition system, which uses an RNN model to map from audio clip x to a text transcript y. Your algorithm uses beam search to try to find the value of y that maximizes  $P(y \mid x)$ .

On a dev set example, given an input audio clip, your algorithm outputs the transcript  $\hat{y}=$  "I'm building an A Eye system in Silly con Valley.", whereas a human gives a much superior transcript  $y^*=$  "I'm building an AI system in Silicon Valley."

According to your model,

$$P(\hat{y} \mid x) = 1.09 * 10^{-7}$$

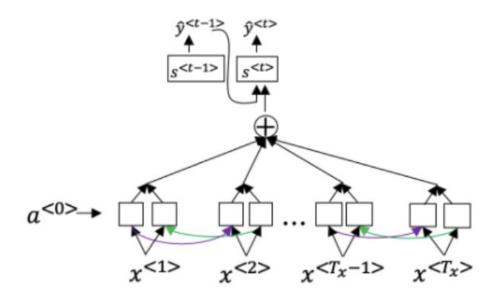
$$P(y^* \mid x) = 7.21 * 10^-8$$

| Would you expect increasing the beam width B to help correct this example?  |  |  |  |  |
|---|--|--|--|--|
| No, because $P(y^*\mid x)\leq P(\hat{y}\mid x)$ indicates the error should be attributed to the RNN rather than to the search algorithm.  |  |  |  |  |
| Correct   |  |  |  |  |
| No, because $P(y^*\mid x)\leq P(\hat{y}\mid x)$ indicates the error should be attributed to the search algorithm rather than to the RNN.  |  |  |  |  |
| Yes, because $P(y^*\mid x)\leq P(\hat{y}\mid x)$ indicates the error should be attributed to the RNN rather than to the search algorithm.   |  |  |  |  |
| Yes, because $P(y^*\mid x)\leq P(\hat{y}\mid x)$ indicates the error should be attributed to the search algorithm rather than to the RNN.   |  |  |  |  |
| Continuing the example from Q4, suppose you work on your algorithm for a few more weeks, and now find that for the vast majority of examples on which your algorithm makes a mistake, $P(y^* \mid x) > P(\hat{y} \mid x)$ . This suggest you should focus your attention on improving the search algorithm. |  |  |  |  |
| True.   |  |  |  |  |
| Correct   |  |  |  |  |

5.

False.

6. Consider the attention model for machine translation.



Further, here is the formula for  $\alpha^{< t, t'>}$ .

$$\alpha^{< t, t'>} = \frac{\exp(e^{< t, t'>})}{\sum_{t'=1}^{T_x} \exp(e^{< t, t'>})}$$

Which of the following statements about  $\alpha^{< t, t'>}$  are true? Check all that apply.



We expect  $\alpha^{< t,t'>}$  to be generally larger for values of  $a^{< t'>}$  that are highly relevant to the value the network should output for  $y^{< t>}$ . (Note the indices in the superscripts.)

## Correct

We expect  $\alpha^{< t,t'>}$  to be generally larger for values of  $a^{< t>}$  that are highly relevant to the value the network should output for  $y^{< t'>}$ . (Note the indices in the superscripts.)

| $\sum_t lpha^{< t, t'>} = 1$ (Note the summation is over $t$ .)   |  |  |  |  |
|---|--|--|--|--|
| Un-selected is correct  |  |  |  |  |
| $\sum_{t'} lpha^{< t, t'>} = 1$ (Note the summation is over $t'$ .)   |  |  |  |  |
| Correct   |  |  |  |  |
| The network learns where to "pay attention" by learning the values $e^{< t,t'>}$ , which are computed using a small neural network:   |  |  |  |  |
| We can't replace $s^{< t-1>}$ with $s^{< t>}$ as an input to this neural network. This is because $s^{< t>}$ depends on $\alpha^{< t,t'>}$ which in turn depends on $e^{< t,t'>}$ ; so at the time we need to evalute this network, we haven't computed $s^{< t>}$ yet. |  |  |  |  |
| True  |  |  |  |  |
| Correct   |  |  |  |  |
| False   |  |  |  |  |
| Compared to the encoder-decoder model shown in Question 1 of this quiz (which does not use an attention mechanism), we expect the attention model to have the greatest advantage when:  |  |  |  |  |
| $lacksquare$ The input sequence length $T_x$ is large.  |  |  |  |  |
| Correct   |  |  |  |  |
| $igcup$ The input sequence length $T_x$ is small.   |  |  |  |  |

7.

8.

| 9.  | Under the CTC model, identical repeated characters not separated by the "blank" character (_) are collapsed. Under the CTC model, what does the following string collapse to? |   |  |  |  |
|---|---|---|--|--|--|
|   | c_oo  | o_o_kkb_oooooookkk  |  |  |  |
|   |   | cokbok  |  |  |  |
|   |   | cookbook  |  |  |  |
|   | Corr  | rect  |  |  |  |
|   |   | cook book   |  |  |  |
|   |   | coookkbooooookkk  |  |  |  |
| $10.$ In trigger word detection, $x^{< t>}$ is: |   |   |  |  |  |
|   |   | Features of the audio (such as spectrogram features) at time $t$ .                        |  |  |  |
|   | Corr  | ect   |  |  |  |
|   |   | The $t\text{-th}$ input word, represented as either a one-hot vector or a word embedding. |  |  |  |
|   |   | Whether the trigger word is being said at time $t. $                                      |  |  |  |
|   |   | Whether someone has just finished saying the trigger word at time $t$ .                   |  |  |  |
|   |   |   |  |  |  |