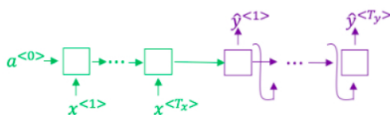


1. Consider using this encoder-decoder model for machine translation.

1 point



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  $x$ .

☒ False

☐ True

44:29

↗ Expand

2. In beam search, if you decrease the beam width  $B$ , which of the following would you expect to be true? Select all that apply.

1 point

☐ Beam search will use up more memory.

☒ Beam search will run more quickly.

☒ Beam search will converge after fewer steps.

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

44:28

↗ Expand

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

1 point

☒ False

☐ True

44:28

↗ Expand

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  $P(y|x)$ .

1 point

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,

$$P(\hat{y} | x) = 1.95 \cdot 10^{-7}$$

$$P(y^* | x) = 3.42 \cdot 10^{-9}$$

True/False: Trying a different network architecture could help correct this example.

☒ True

☐ False

44:27

[Expand](#)

5. 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,  $P(y^* | x) > P(\hat{y} | x)$ . This suggests you should not focus your attention on improving the search algorithm.

1 point

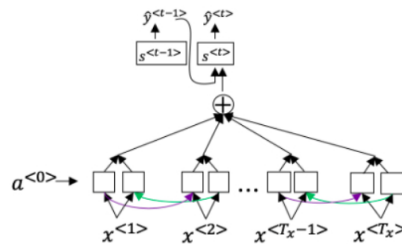
☐ True☒ False

44:27

[Expand](#)

6. Consider the attention model for machine translation.

1 point



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.)

☐  $\alpha^{<t, t'>}$  is equal to the amount of attention  $y^{<t>}$  should pay to  $a^{<t'>}$

☒  $\sum_{t'} \alpha^{<t, t'>} = 1$

☐  $\sum_{t'} \alpha^{<t, t'>} = 0$

44:26

[Expand](#)

7. The network learns where to "pay attention" by learning the values  $e^{<t, t'>}$ , which are computed using a small neural network:

1 point

We can replace  $s^{<t-1>}$  with  $s^{<t>}$  as an input to this neural network because  $s^{<t>}$  is independent of  $\alpha^{<t, t'>}$  and  $e^{<t, t'>}$ .

☐ True☒ False

44:25

[Expand](#)

8. 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:

1 point

- ☐ The input sequence length  $T_x$  is small.
- ☒ The input sequence length  $T_x$  is large.

44:25

[Expand](#)

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?

1 point

\_\_c\_\_oo\_\_o\_\_kk\_\_b\_\_ooooo\_\_oo\_\_kkk

- ☒ cookbook
- ☐ cookkkboooooookkk
- ☐ cokbok
- ☐ cook book

44:25

[Expand](#)

10. In trigger word detection,  $x^{<t>}$  is:

1 point

- ☐ Whether someone has just finished saying the trigger word at time  $t$ .
- ☐ Whether the trigger word is being said at time  $t$ .
- ☐ The  $t$ -th input word, represented as either a one-hot vector or a word embedding.
- ☒ Features of the audio (such as spectrogram features) at time  $t$ .

44:24

[Expand](#)