

Encoder-decoder architectures

LATEST SUBMISSION GRADE

100%

1.Question 1

In the lecture as well as in this test we will have lots of formulas. Let us first make sure that we remember the used notation.

Please, name the following objects: $I, J, x_i, y_j, h_i, v_j, s_j$

- ☐ length of target, length of source, encoder state, decoder state, source word, target word, context vector
- ☐ length of source, length of target, source word, target word, encoder state, decoder state, context vector
- ☒ length of source, length of target, source word, target word, encoder state, context vector, decoder state

Correct

1 / 1 point

2.Question 2

How do we compute the context (thought) vector v for the decoder position j in a seq2seq model without attention?

- ☒ h_I , where h_I is the last encoder state
- ☐ h_j , where h_j is the j -th state of the encoder
- ☐ $\sum_i \alpha_i h_i$, where α_i are some weights

Correct

Correct!

1 / 1 point

3.Question 3

How many new parameters for the network are introduced to calculate **multiplicative attention** weights? (Just to calculate, we are not yet looking into how we use them afterwards).

- ☐ The length of the source, multiplied by the length of the target
- ☒ The dimension of an encoder state, multiplied by the dimension of a decoder state
- ☐ No new parameters

Correct

Exactly! This is the number of parameters for **multiplicative** attention.

1 / 1 point

4. Question 4

Which of the following formulas stand for the **additive attention**? Note that h_i is the i -th encoder state, s_j is the j -th decoder state, and we are interested in the similarity between them.

- ☐ $h_i^T W s_j$, where W is a matrix of parameters
- ☐ $h_i^T s_j$
- ☒ $w^T \tanh(W [h_i, s_j])$, where the brackets denote concatenation of the vectors, and w and W are a vector and a matrix of parameters respectively.

Correct

Exactly! Take a moment to see, that this is just a different form of the same additive attention formula that was introduced at the lecture.

1 / 1 point

5. Let us denote encoder states by h_i with i going from 1 to I . Let us denote by α_i^j the similarities computed using the additive attention formula from the previous question. How should we compute the context vector v_j for the decoder position j ?

1 / 1 point

- ☒ $\sum_{i=1}^I \frac{\exp \alpha_i^j}{\sum_{i'} \exp \alpha_{i'}^j} h_i$
- ☐ $\sum_{i=1}^I \frac{\exp \alpha_i^j}{\sum_{j'} \exp \alpha_{i'}^j} h_i$
- ☐ $\sum_{i=1}^I \alpha_i^j h_i$

✓ **Correct**

Correct! We apply *softmax* to transform the weights into probabilities and compute the average of the encoder states.

1 / 1 point

6. Question 6

Which three vectors should be passed to a decoder state s_j in a seq2seq with attention model from the lecture?

- ☐ h_{i-1} - the previous encoder state
- ☐ h_i - the i -th encoder state
- ☐ y_j - the j -th word in the target sequence
- ☐ x_{i-1} - the previous word in the source sequence
- ☒ y_{j-1} - the previous word in the target sequence

Correct

- ☒ s_{j-1} - the previous decoder state

Correct

- ☐ s_{js} - the jj -th decoder state
- ☒ v_{jv} - the context vector for position jj , calculated using attention

Correct

- ☐ x_{ix} - the ii -th word in the source sequence

1 / 1 point

7.Question 7

Which techniques would help if the data has rich morphology, informal spelling, and other sources of OOV tokens?

- ☐ Negative sampling
- ☒ Copy mechanism

Correct

- ☒ Byte-pair encoding

Correct

- ☒ Sub-word modeling

Correct

- ☐ Hierarchical softmax

1 / 1 point

8.Question 8

Let us imagine we have trained a conversational chat-bot as a seq2seq model on Harry Potter movies subtitles. What problems could we expect?

- ☒ When asked "What's your name?", the bot is not sure and says Harry, or Ron, or Hermione from time to time.

Correct

No personality problem!

- ☒ The bot doesn't remember what has already been decided in your dialogue

Correct

No memory of the context / history!

- ☒ The bot suggests to use a time-turner or probably some spell if you say you do not have enough time for your Coursera studies

Correct

Dramatic/unrealistic topics, influenced by the training corpus!

- ☐ The bot makes lots of spelling mistakes