

Your grade: 100%**Next item →**

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1. A Transformer Network, unlike its predecessors RNNs, GRUs and LSTMs, can process entire sentences all at the same time. (Parallel architecture).

1 / 1 point

 True False**Expand****Correct**

A Transformer Network can ingest entire sentences all at the same time.

2. Transformer Network methodology is taken from:

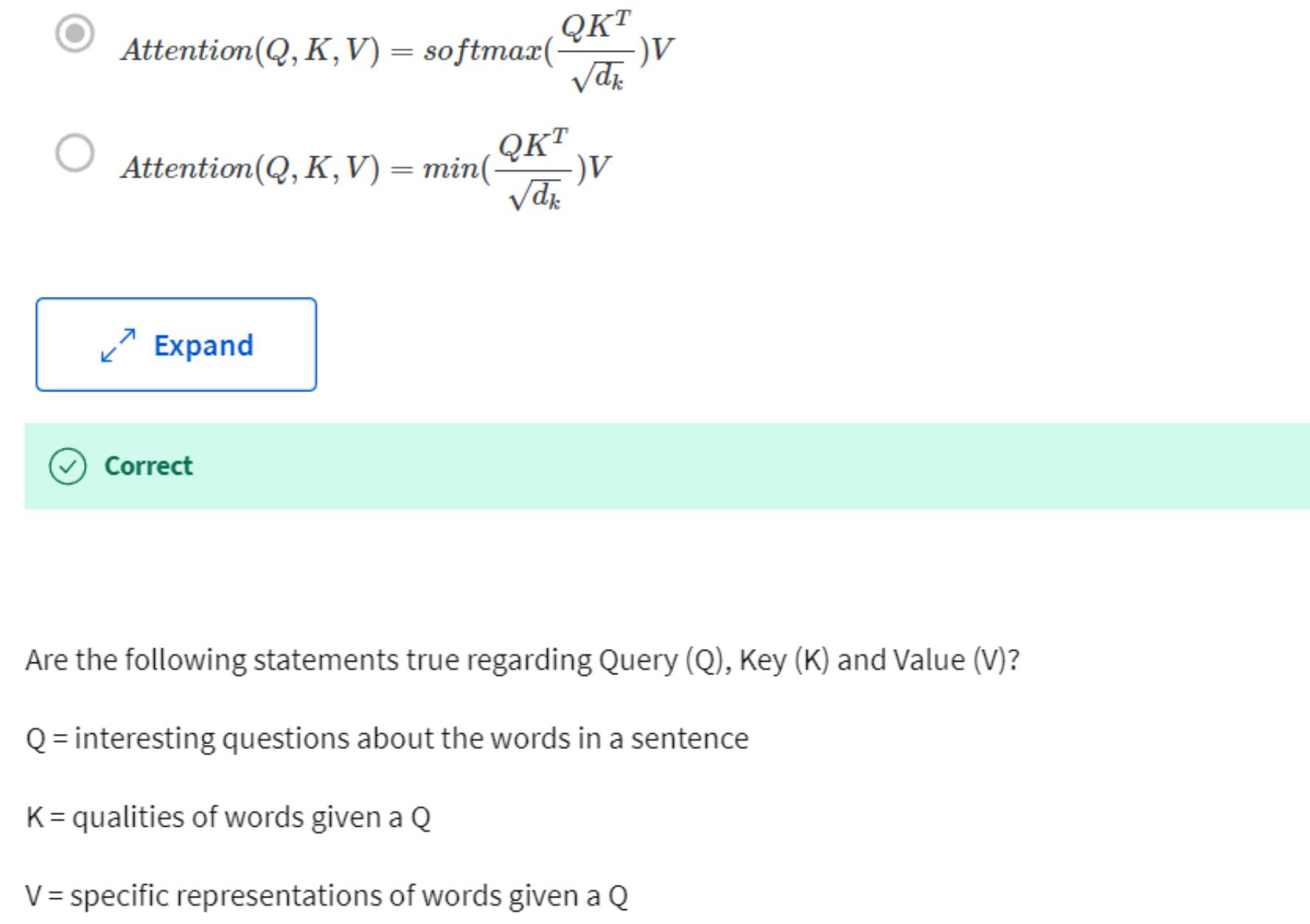
1 / 1 point

 RNN and LSTMs Attention Mechanism and RNN style of processing. GRUs and LSTMs Attention Mechanism and CNN style of processing.**Expand****Correct**

Transformer architecture combines the use of attention based representations and a CNN convolutional neural network style of processing.

3. How does the Self-Attention mechanism of transformers use neighboring words to compute a word's context?

1 / 1 point

 Summation of the word values to map the Attention related to that given word. Multiplication of the word values to map the Attention related to that given word. Selecting the minimum word values to map the Attention related to that given word. Selecting the maximum word values to map the Attention related to that given word.**Expand****Correct**

Given a word, its neighboring words are used to compute its context by summing up the word values to map the Attention related to that given word.

4. Which of the following correctly represents *Attention*?

1 / 1 point

 $\text{Attention}(Q, K, V) = \min\left(\frac{QV^T}{\sqrt{d_k}}\right)K$ $\text{Attention}(Q, K, V) = \text{softmax}\left(\frac{QK^T}{\sqrt{d_k}}\right)V$ $\text{Attention}(Q, K, V) = \min\left(\frac{QK^T}{\sqrt{d_k}}\right)V$ **Expand****Correct**

5. Are the following statements true regarding Query (Q), Key (K) and Value (V)?

1 / 1 point

Q = interesting questions about the words in a sentence

K = qualities of words given a Q

V = specific representations of words given a Q

 False True**Expand****Correct**

Q = interesting questions about the words in a sentence, K = qualities of words given a Q, V = specific representations of words given a Q

Attention($W_i^Q Q, W_i^K K, W_i^V V$)

1 / 1 point

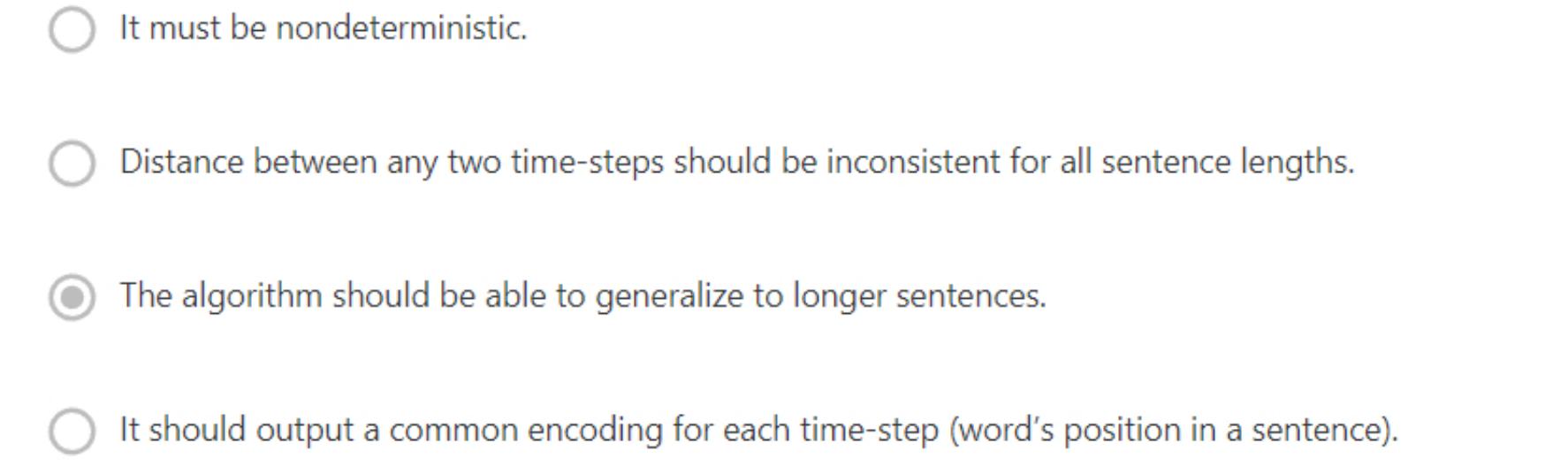
6. i here represents the computed attention weight matrix associated with the *i*th "head" (sequence).

 True False**Expand****Correct**

Great, you got all the right answers.

7. Following is the architecture within a Transformer Network (*without displaying positional encoding and output layers(s)*).

1 / 1 point

What information does the *Decoder* take from the *Encoder* for its second block of *Multi-Head Attention*? (Marked *X*, pointed by the independent arrow)

(Check all that apply)

 V**Correct** K**Correct** Q**Expand****Correct**

Great, you got all the right answers.

8. Following is the architecture within a Transformer Network. (*without displaying positional encoding and output layers(s)*)

1 / 1 point

What is the output layer(s) of the *Decoder*? (Marked *Y*, pointed by the independent arrow) Softmax layer Linear layer followed by a softmax layer. Softmax layer followed by a linear layer. Linear layer**Expand****Correct**

This is a good criterion for a good positional encoding algorithm.

10. Which of these is a good criterion for a good positional encoding algorithm?

1 / 1 point

 It must be nondeterministic. Distance between any two time-steps should be inconsistent for all sentence lengths. The algorithm should be able to generalize to longer sentences. It should output a common encoding for each time-step (word's position in a sentence).**Expand****Correct**

Positional encoding allows the transformer network to offer an additional benefit over the attention model.