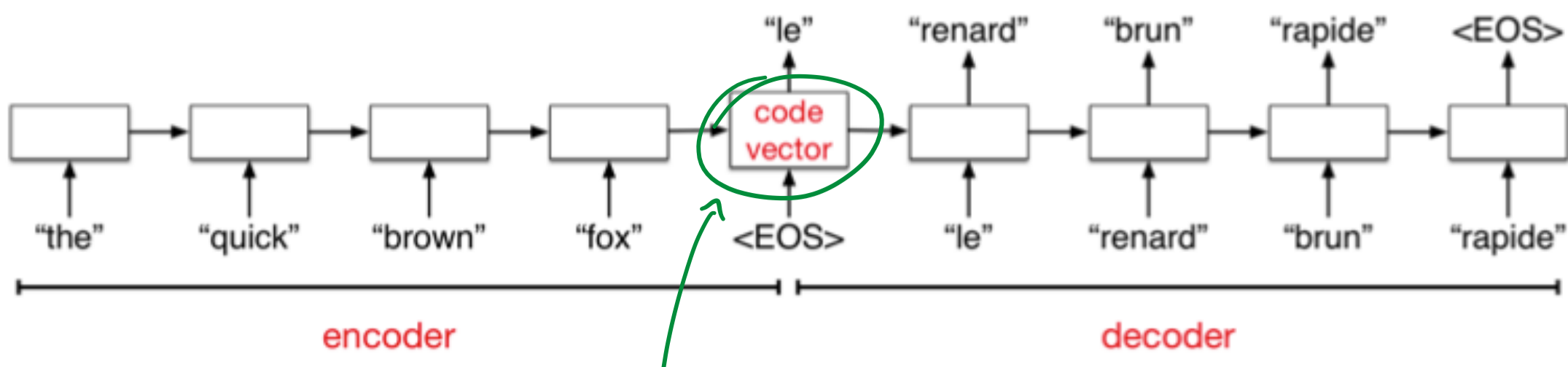


Original encoder-decoder architecture for neural machine translation



memorize all relevant information about the input sentence

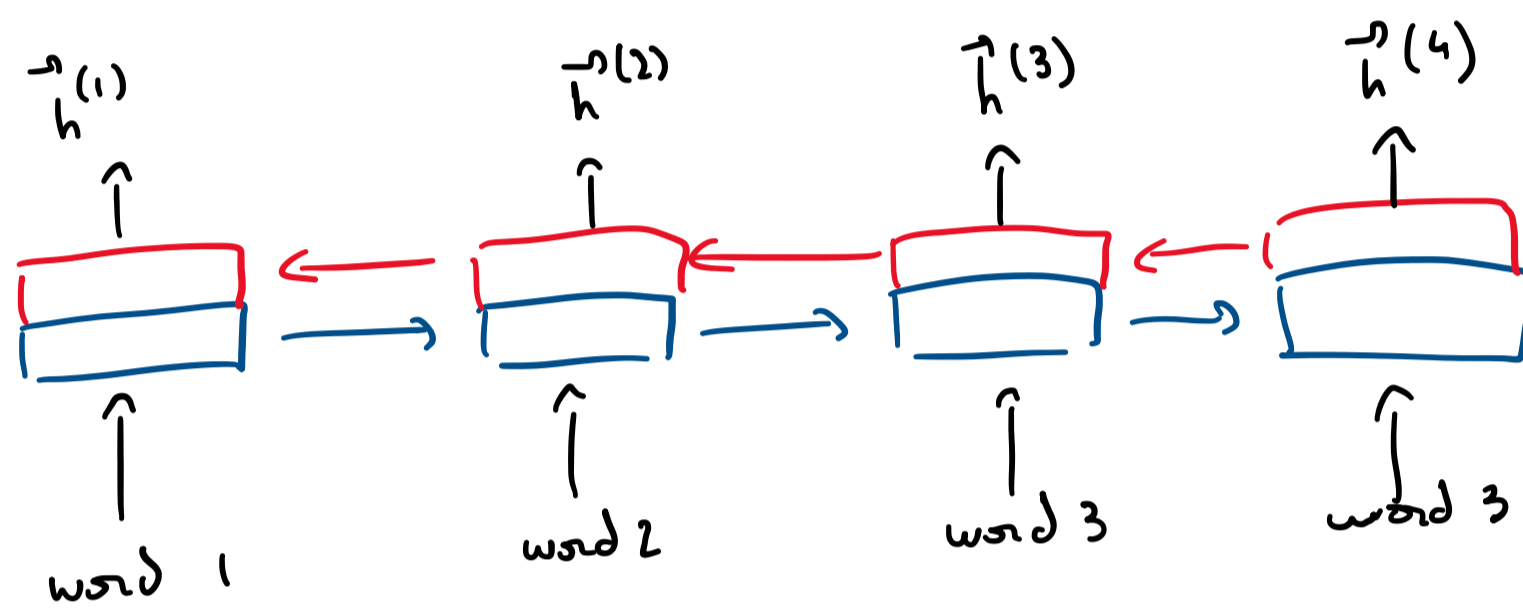
Attention-based modeling

→ look at input sequence (sentence) when generating next

Encoder: bidirectional RNN.

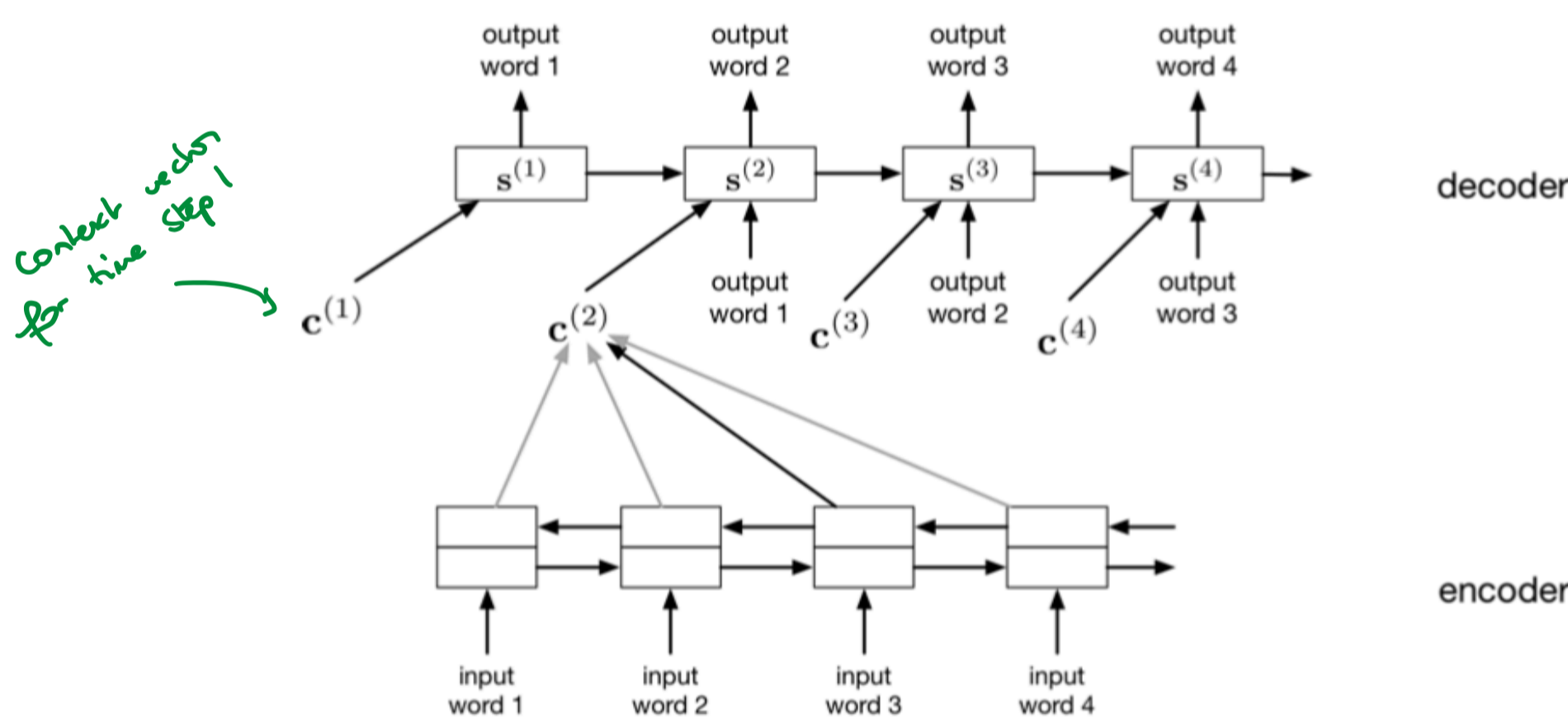
Goal: compute an annotation vector (which the decoder will consult when it generates the output sequence)

Bidirectional: two RNNs
 ↳ one processes the words in the forward order
 ↳ second processes the words in the backward order



Annotation vector: concatenation of the hidden units of the 2 RNNs.

Decoder: will get a context vector in addition to the words (generated up to current time step) as its inputs



Soft-attention model:

$\vec{c}^{(i)}$: context vector for output time step i (decoder)
 t : index of time steps in the input (encoder)

$$\vec{c}^{(i)} = \sum_t \alpha_{it} \vec{h}^{(t)}$$

attention weight

$$\alpha_{it} = \frac{e^{z_{it}}}{\sum_t e^{z_{it}}}$$

previous decoder hidden state
 $z_{it} = a(\vec{s}^{(i-1)}, \vec{h}^{(t)})$
 feed forward neural network
 annotation vector (used for context-based addressing)

