

Sequence to Sequence Models

Deep Learning

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Outline

- Basic models
- Beam search
- Attention mechanism



Basic models



Sequence to Sequence Model

입력/출력

같이 봐주세요



$x^{<1>}$ $x^{<2>}$ $x^{<3>}$ $x^{<4>}$ $x^{<5>}$

Jane visite l'Afrique en septembre

Jane is visiting Africa in September

$y^{<1>}$ $y^{<2>}$ $y^{<3>}$ $y^{<4>}$ $y^{<5>}$ $y^{<6>}$

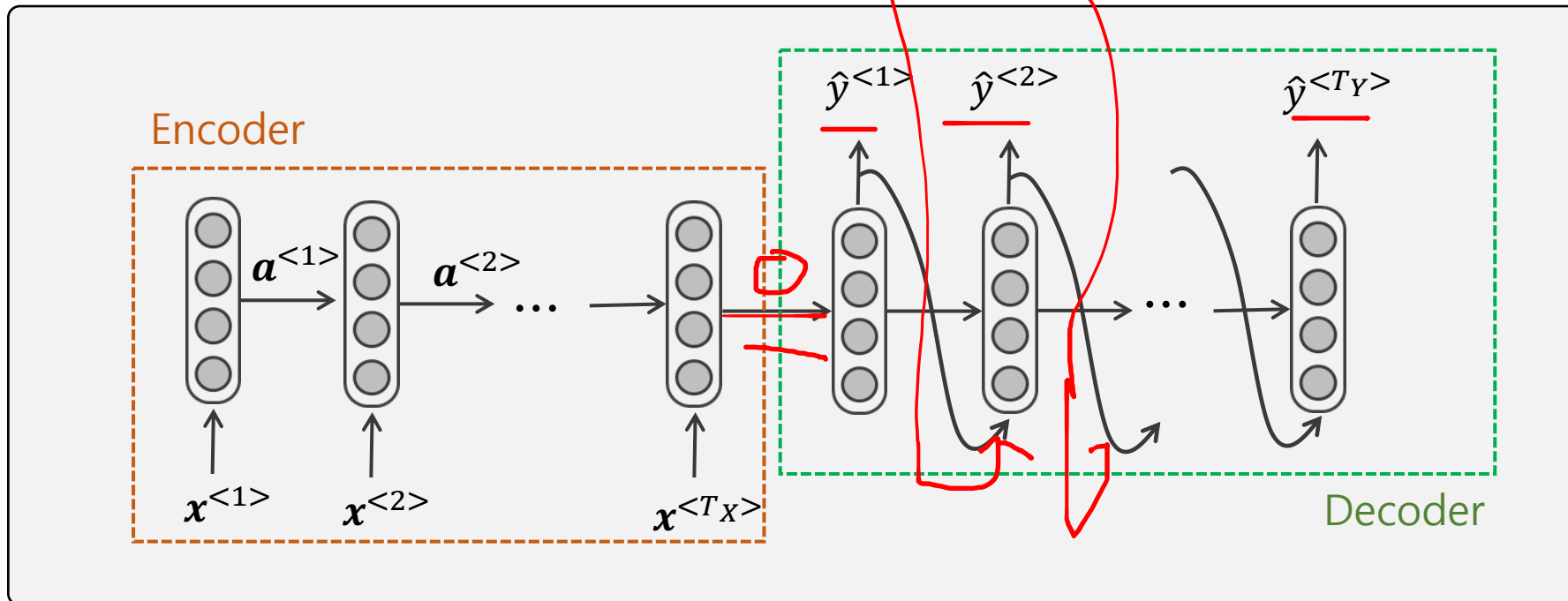
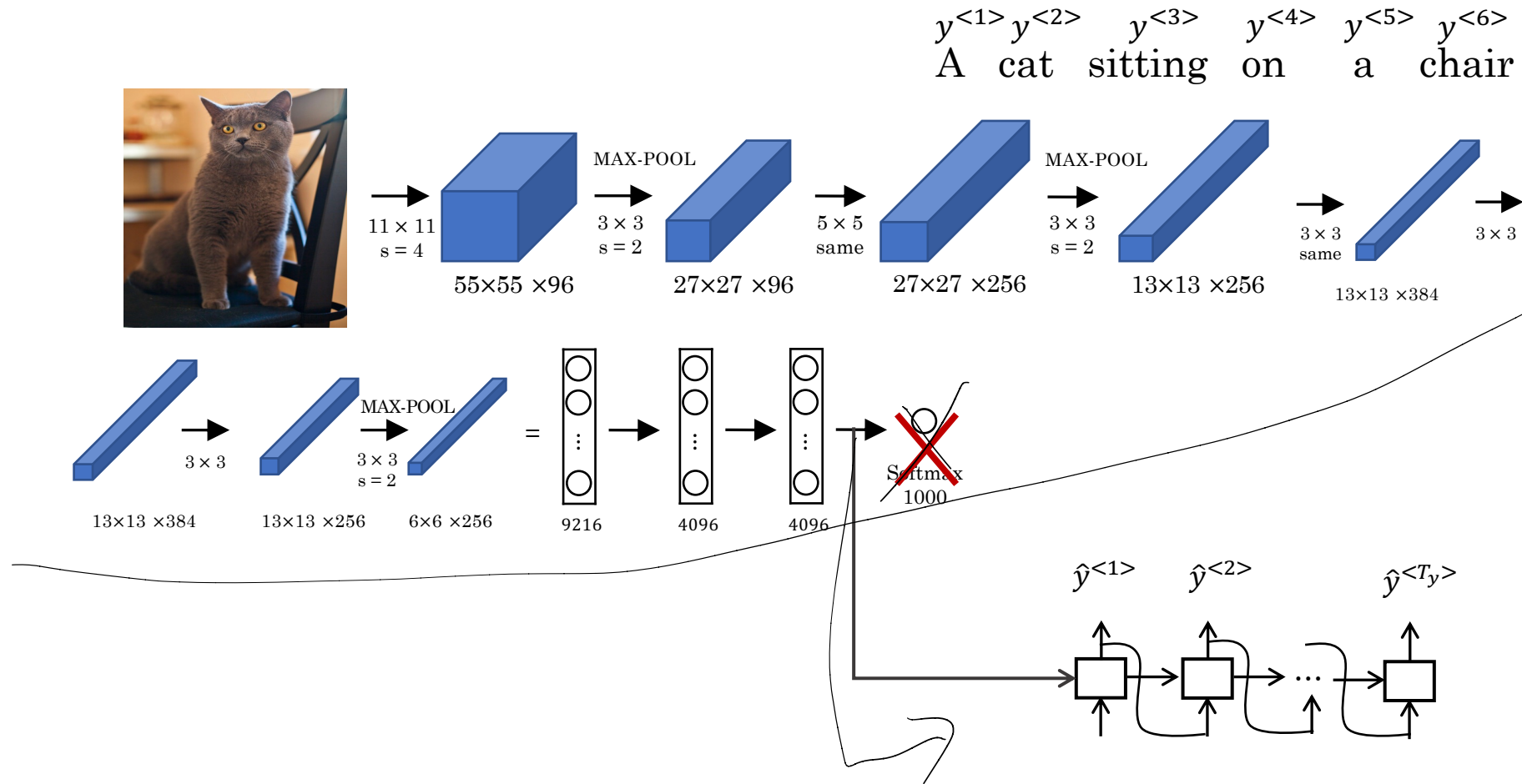


Image Captioning

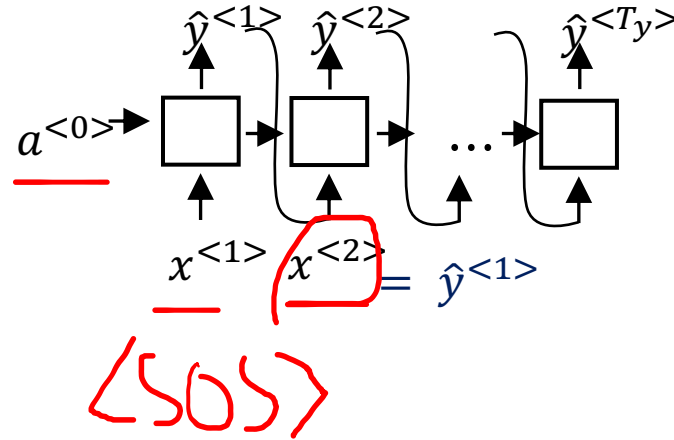
CNN은 이미지 인식용

RNN은 문장 생성



Machine Translation as Building a Conditional Language Model

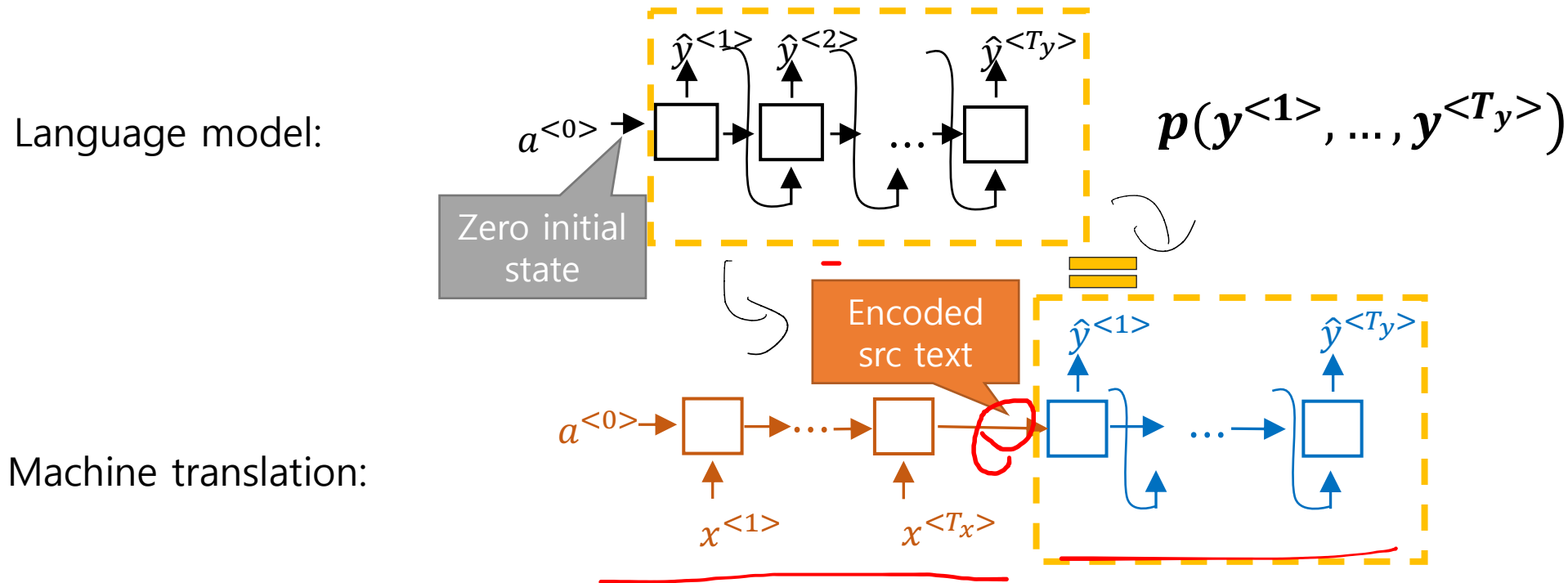
Language model:



$$p(\mathbf{y}^{<1>}, \dots, \mathbf{y}^{<T_y>})$$



Machine Translation as Building a Conditional Language Model



Conditional language model

$$p(y^{<1>}, \dots, y^{<T_y>} \mid x^{<1>}, \dots, x^{<T_x>})$$



Finding the Most likely Translation

Jane visite l'Afrique en septembre.

$$\frac{P(y^{<1>}, \dots, y^{<T_y>} | x)}{\text{English}} \quad \begin{array}{c} \nearrow \\ \text{French} \end{array}$$

- Jane is visiting Africa in September.
- Jane is going to be visiting Africa in September.
- In September, Jane will visit Africa.
- Her African friend welcomed Jane in September.

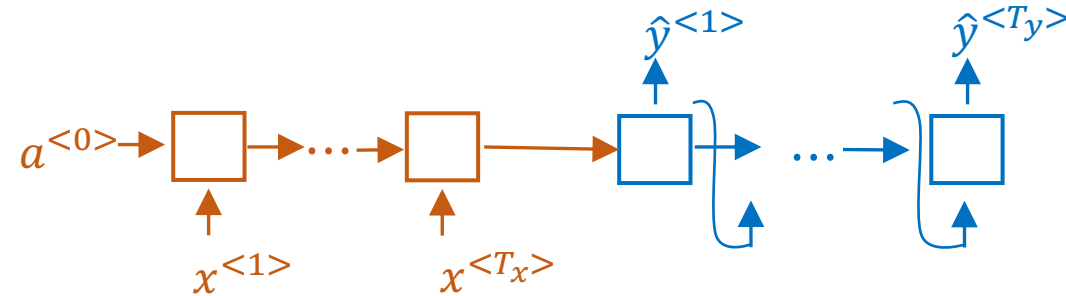
$$\arg \max_{y^{<1>}, \dots, y^{<T_y>}} P(\hat{y}^{<1>}, \hat{y}^{<2>}, \dots, y^{<T_y>} | x)$$



Attention mechanism

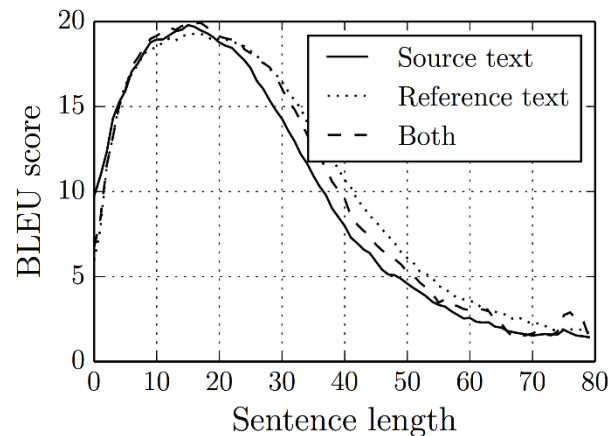


The Problems of Long Sequences



Jane s'est rendue en Afrique en septembre dernier, a apprécié la culture et a rencontré beaucoup de gens merveilleux; elle est revenue en parlant comment son voyage était merveilleux, et elle me tente d'y aller aussi.

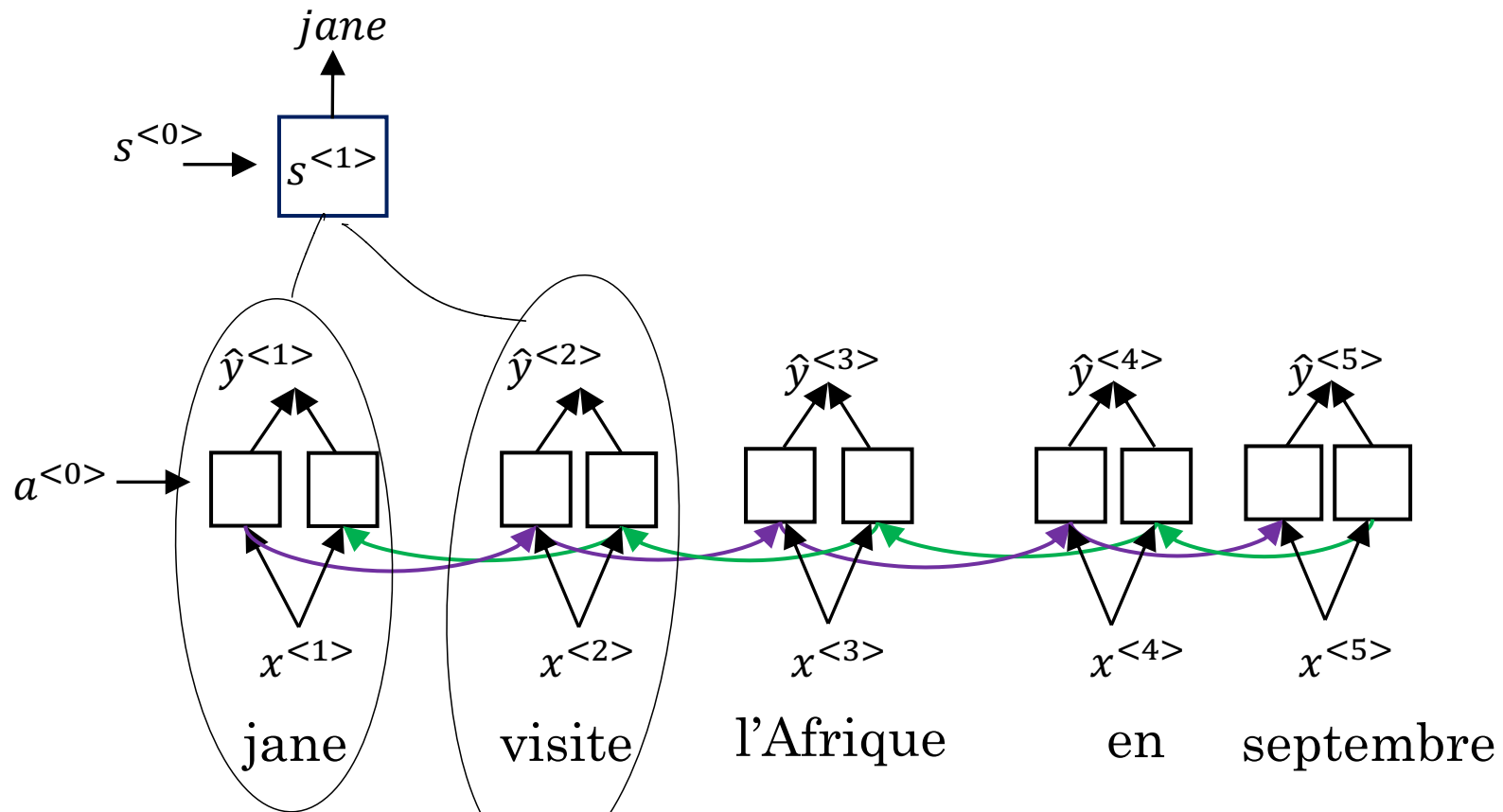
Jane went to Africa last September, and enjoyed the culture and met many wonderful people; she came back raving about how wonderful her trip was, and is tempting me to go too.



https://devblogs.nvidia.com/introduction-neural-machine-translation-gpus-part-3/figure1_bleuscore_vs_sentencelength/



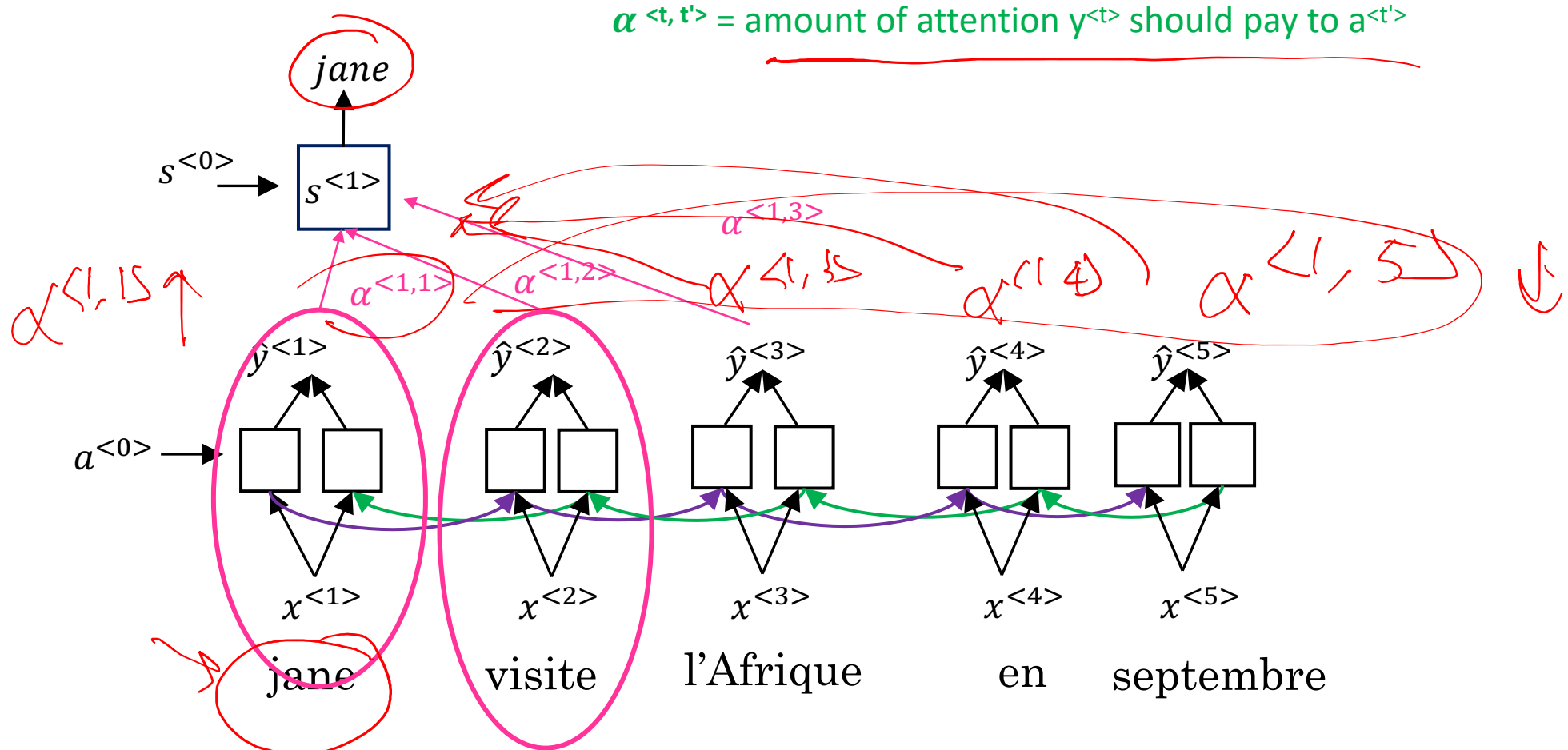
Attention Model intuition



[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



Attention Model intuition

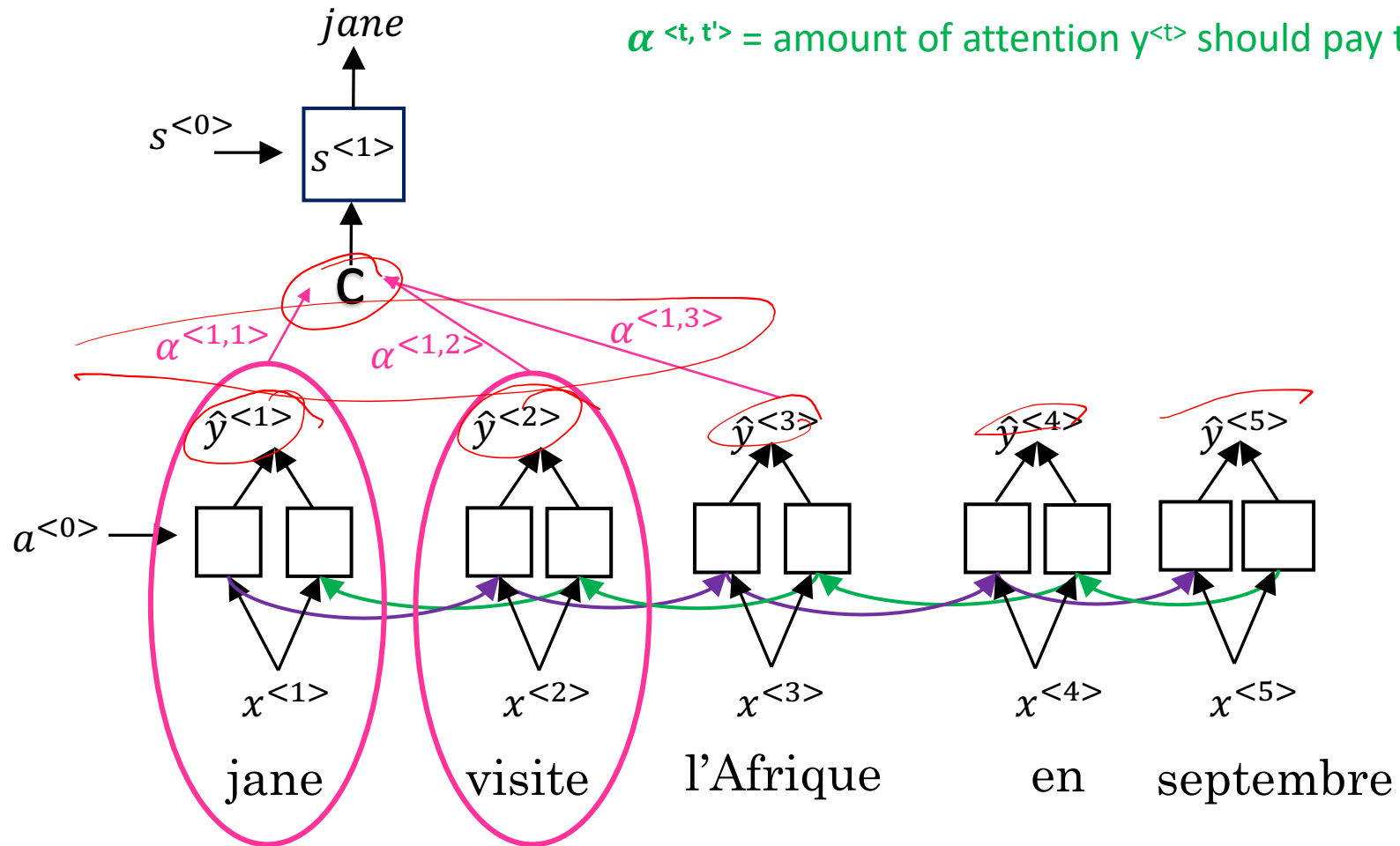


[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]

가장 쉬운 문장 (영문 많은 것)



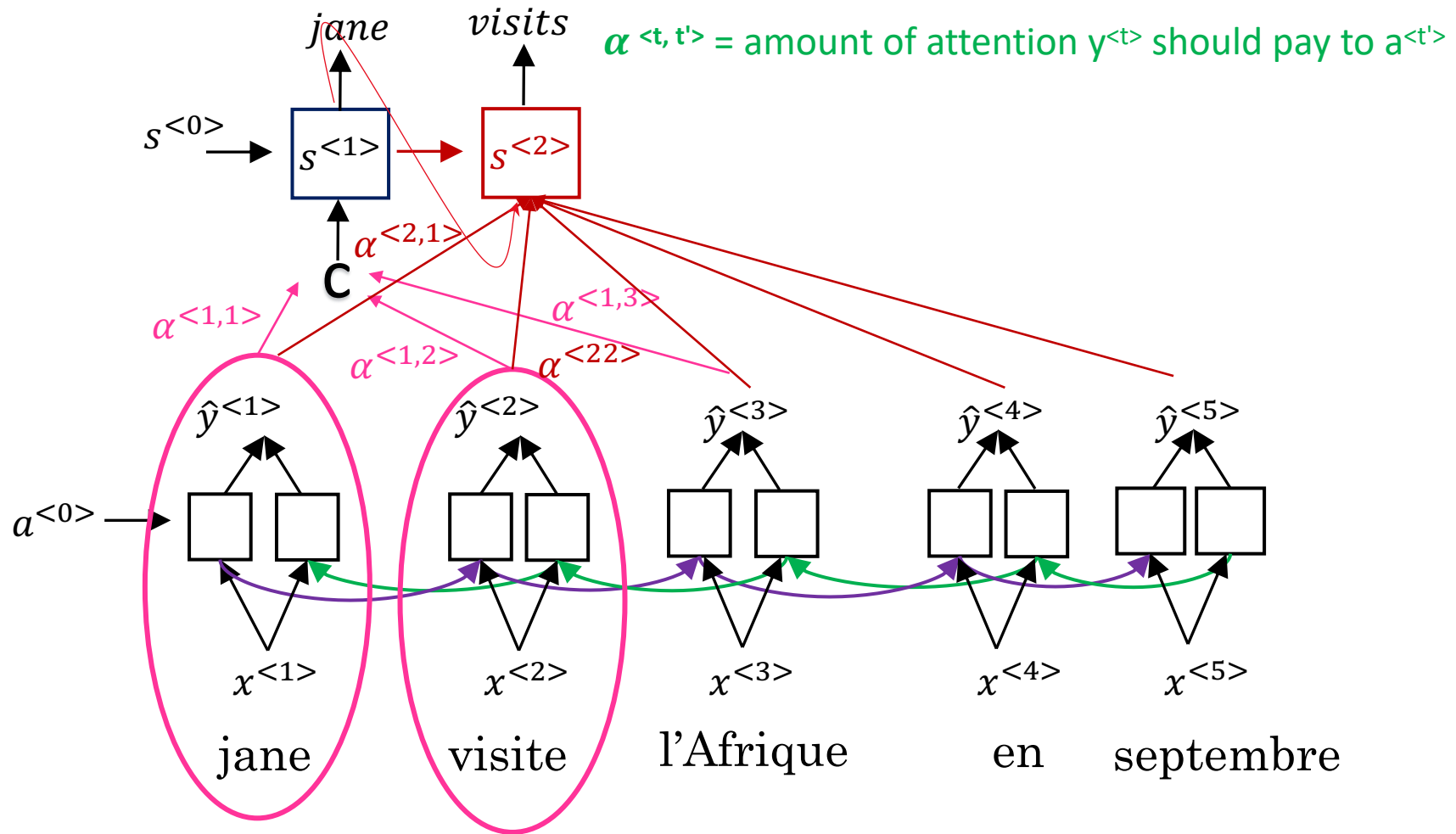
Attention Model intuition



[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



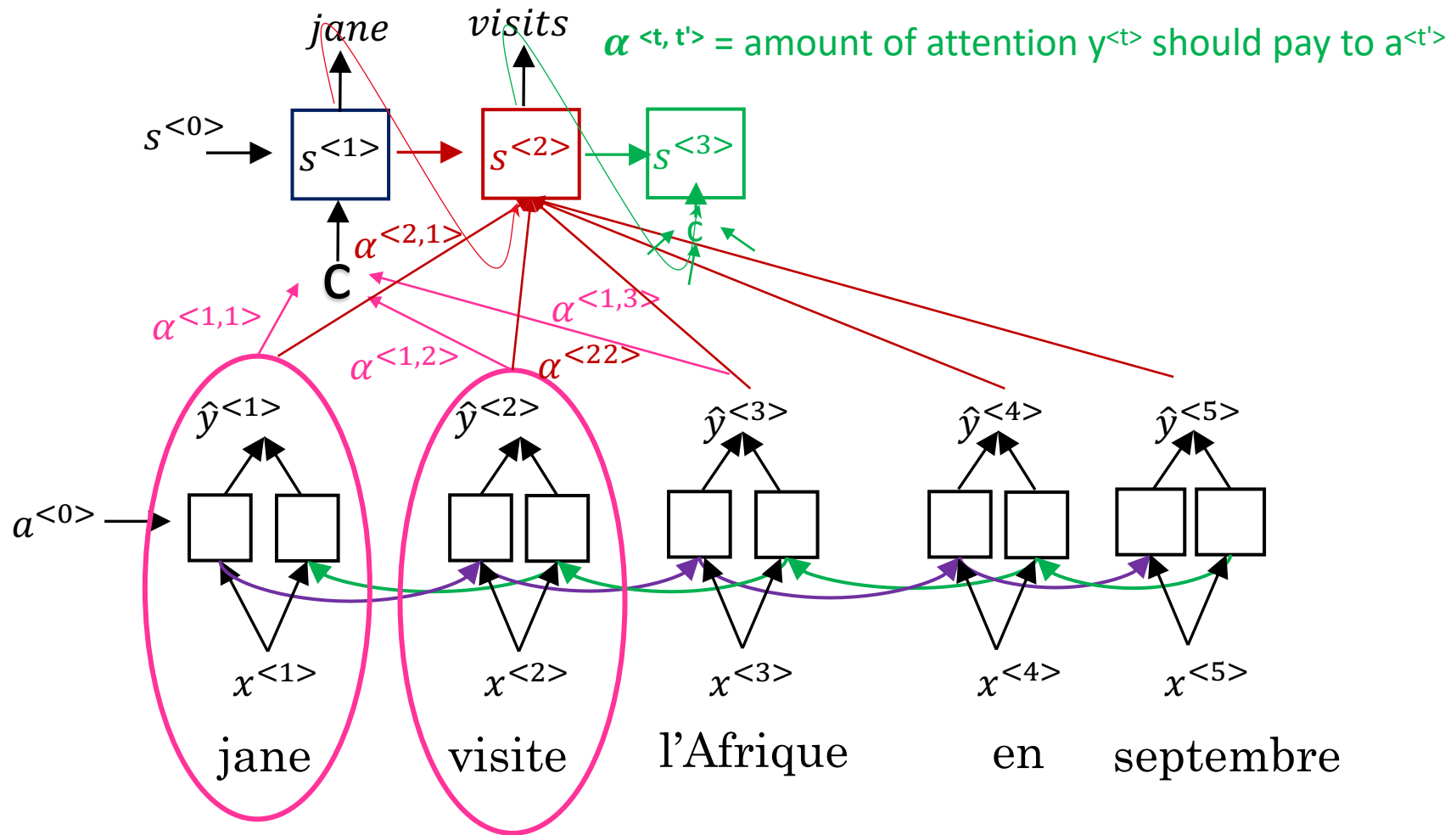
Attention Model intuition



[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



Attention Model intuition

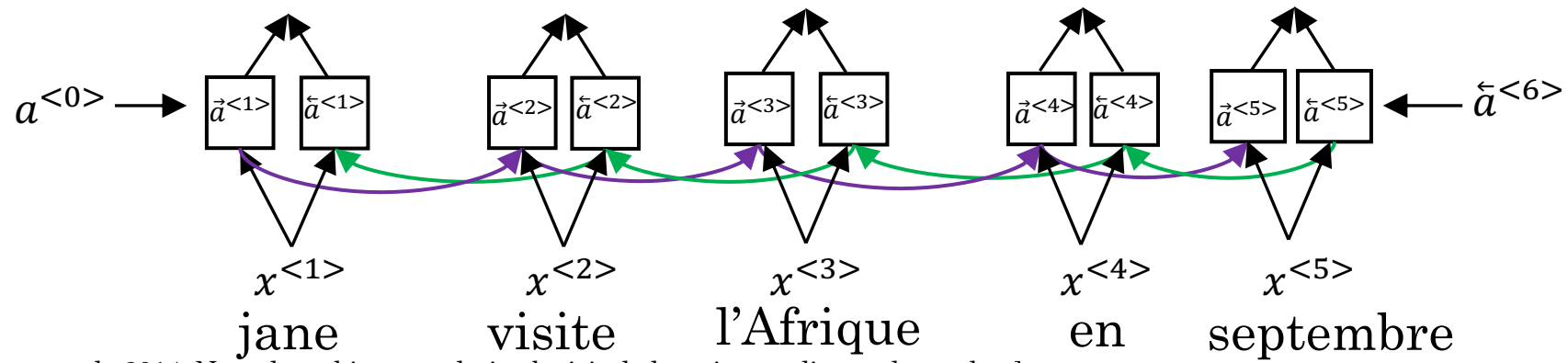


[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



Attention Model

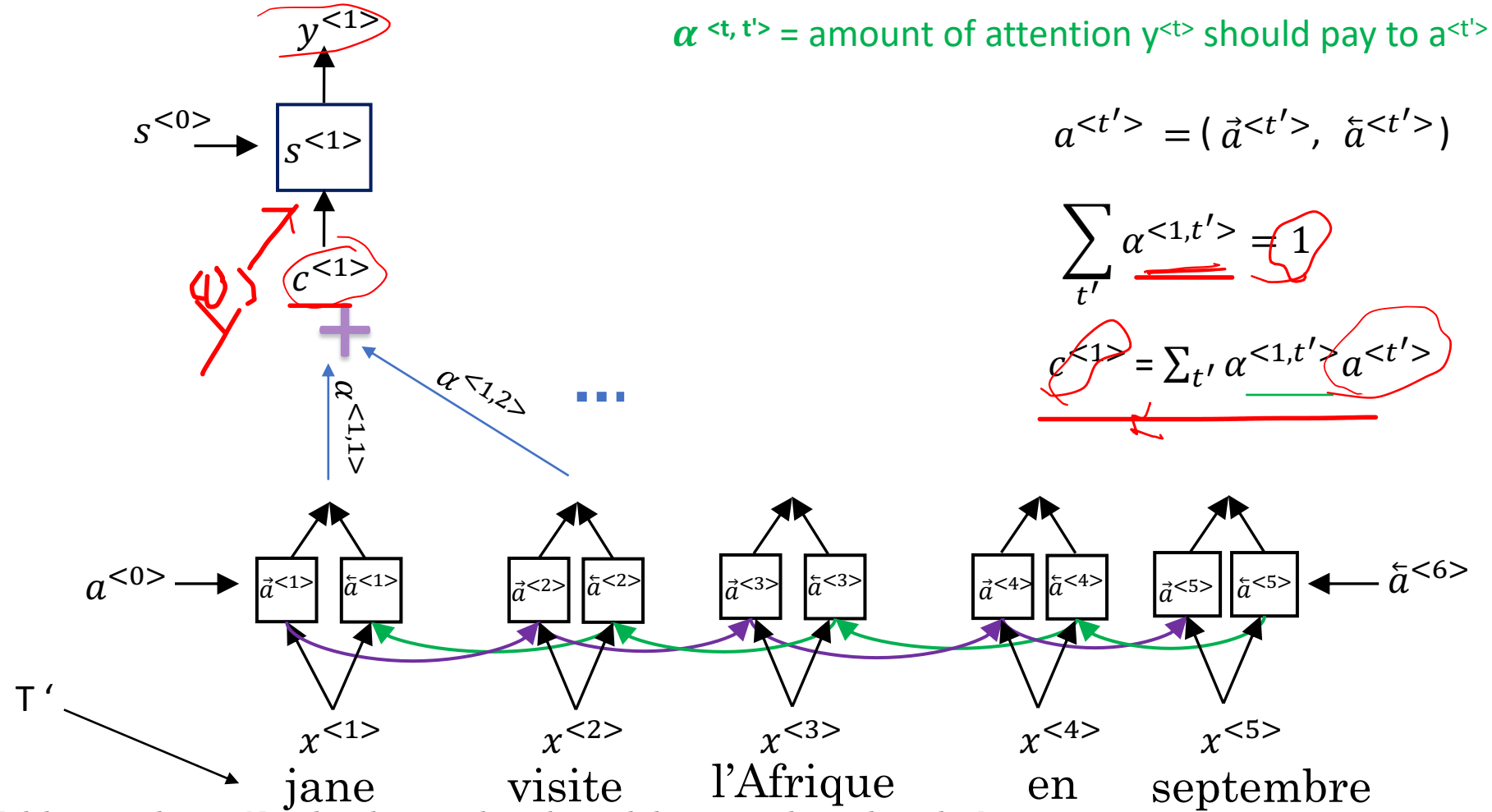
$$a^{<t>} = (\vec{a}^{<t>}, \tilde{a}^{<t>})$$



[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



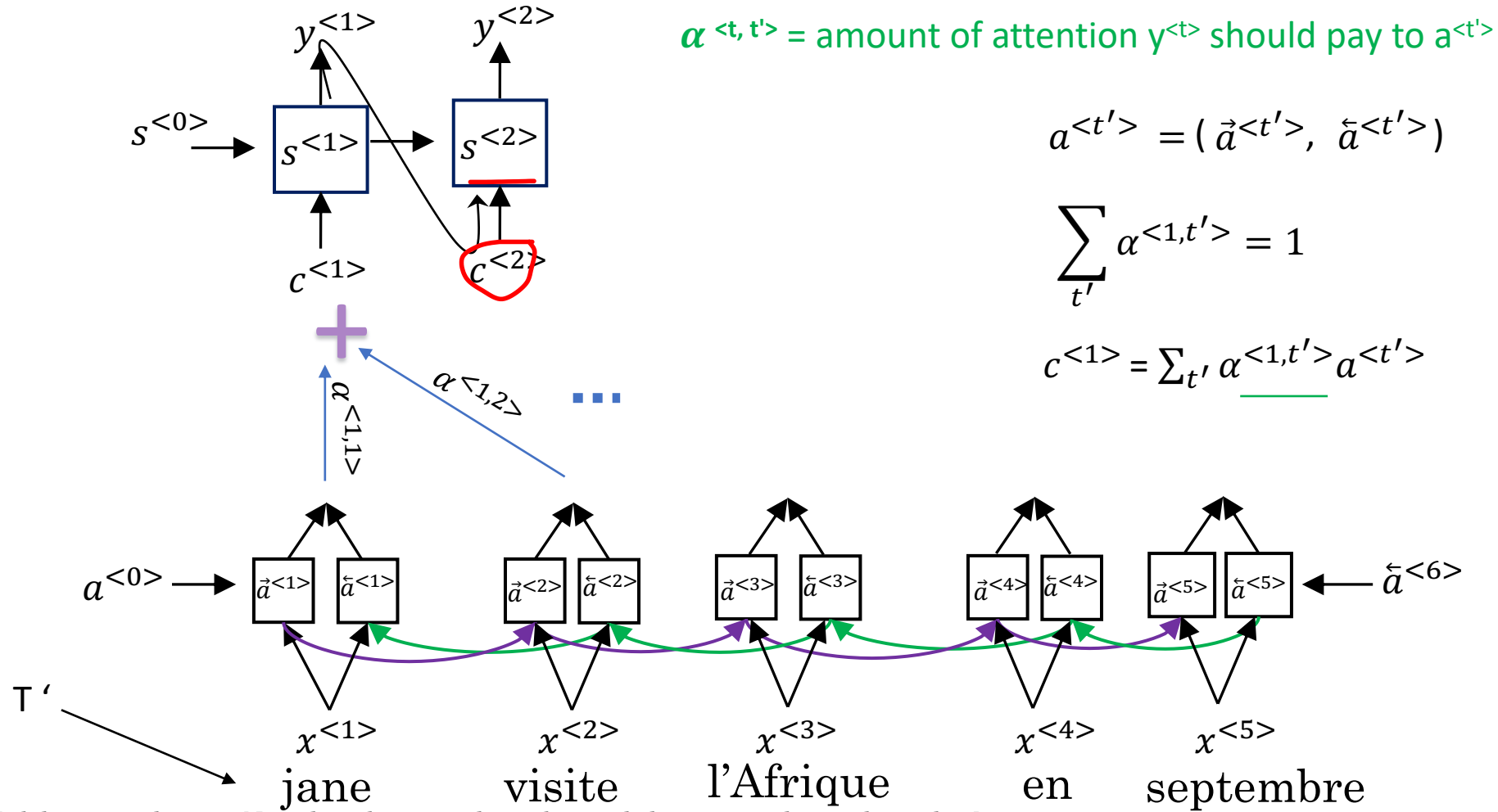
Attention Model



[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



Attention Model

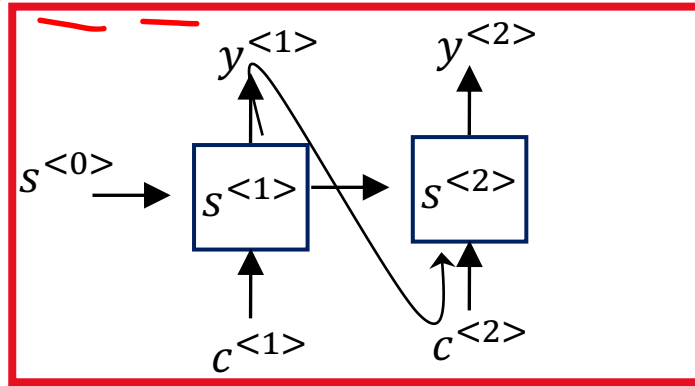


[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



Attention Model

$$s_t = \text{RNNCell}(s_{t-1}, y_{t-1}, c_t)$$

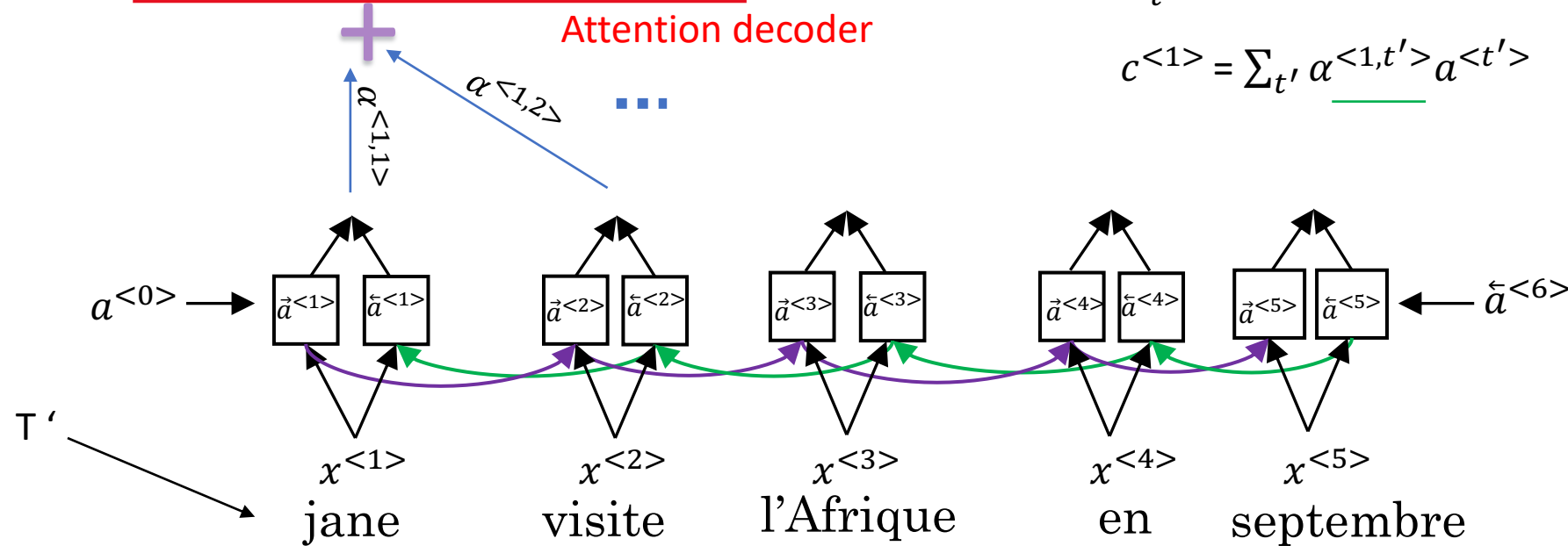


$\alpha^{<t,t'>}$ = amount of attention $y^{<t>}$ should pay to $a^{<t'>}$

$$a^{<t'>} = (\vec{a}^{<t'>}, \tilde{a}^{<t'>})$$

$$\sum_{t'} \alpha^{<1,t'>} = 1$$

$$c^{<1>} = \sum_{t'} \alpha^{<1,t'>} \underline{a^{<t'>}}$$



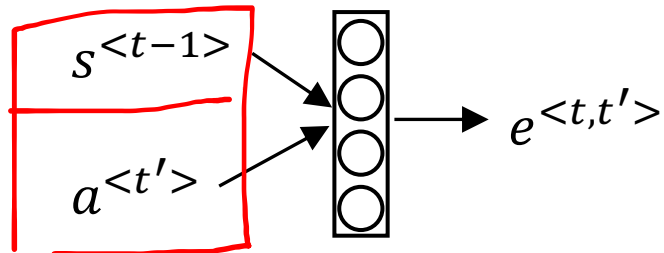
[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



Computing Attention $\alpha^{<t,t'>}$

$\alpha^{<t,t'>}$ = amount of attention $y^{<t>}$ should pay to $a^{<t'>}$

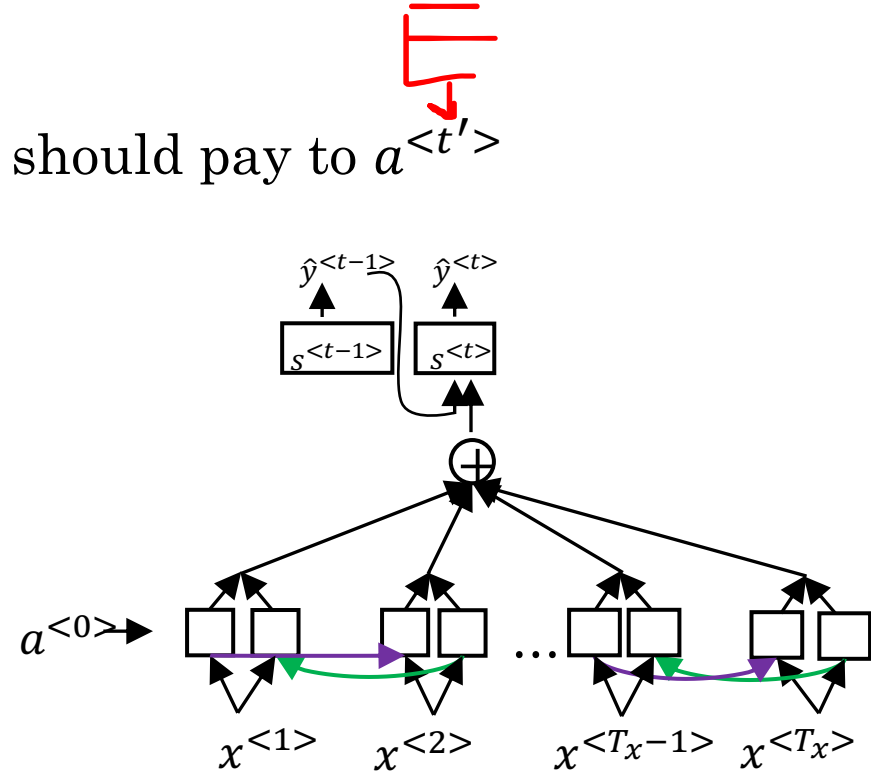
$$\alpha^{<t,t'>} = \frac{\exp(e^{<t,t'>})}{\sum_{t'=1}^{T_x} \exp(e^{<t,t'>})}$$



$$e^{<t,t'>} = v_a^T \tanh(W_{as}s^{<t-1>} + U_{aa}a^{<t'>})$$

$$= v_a^T \tanh(W_a[s^{<t-1>}; a^{<t'>}])$$

[Bahdanau et. al., 2014. Neural machine translation by jointly learning to align and translate]



scala

24 이 어 15

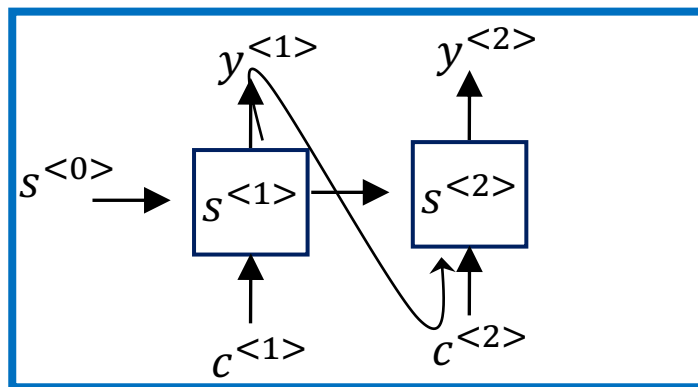


Attention Model

C_t 값은

평균화한 25도

Attention decoder



$$e^{<t,t'>} = v_a^T \tanh(W_a [s^{<t-1>}; a^{<t'>}])$$

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

$$c^{<t>} = \sum_{t'} \alpha^{<t,t'>} a^{<t'>}$$

$$s_t = \text{RNNCell}(s_{t-1}, y_{t-1}, c_t)$$

$C^{<t>}$

제곱

Attention weight가

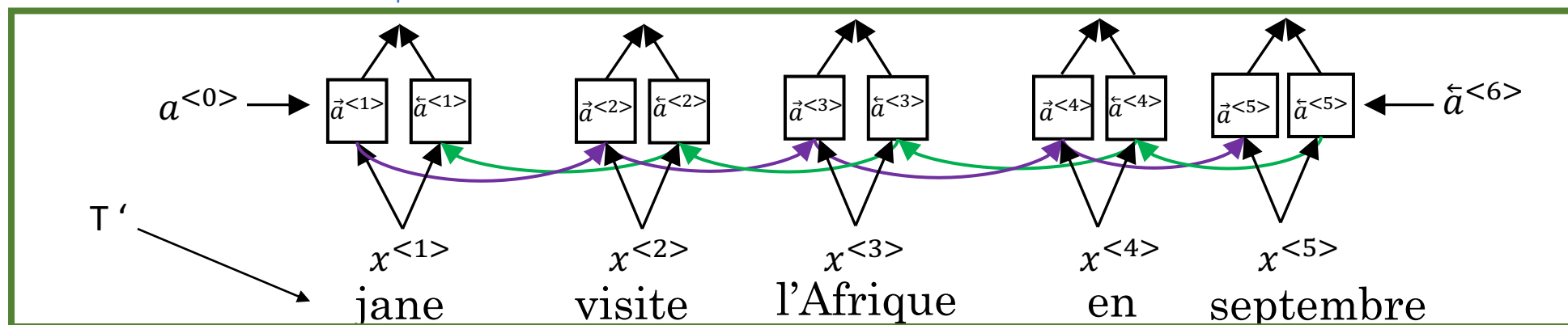
제곱에

큰 영향

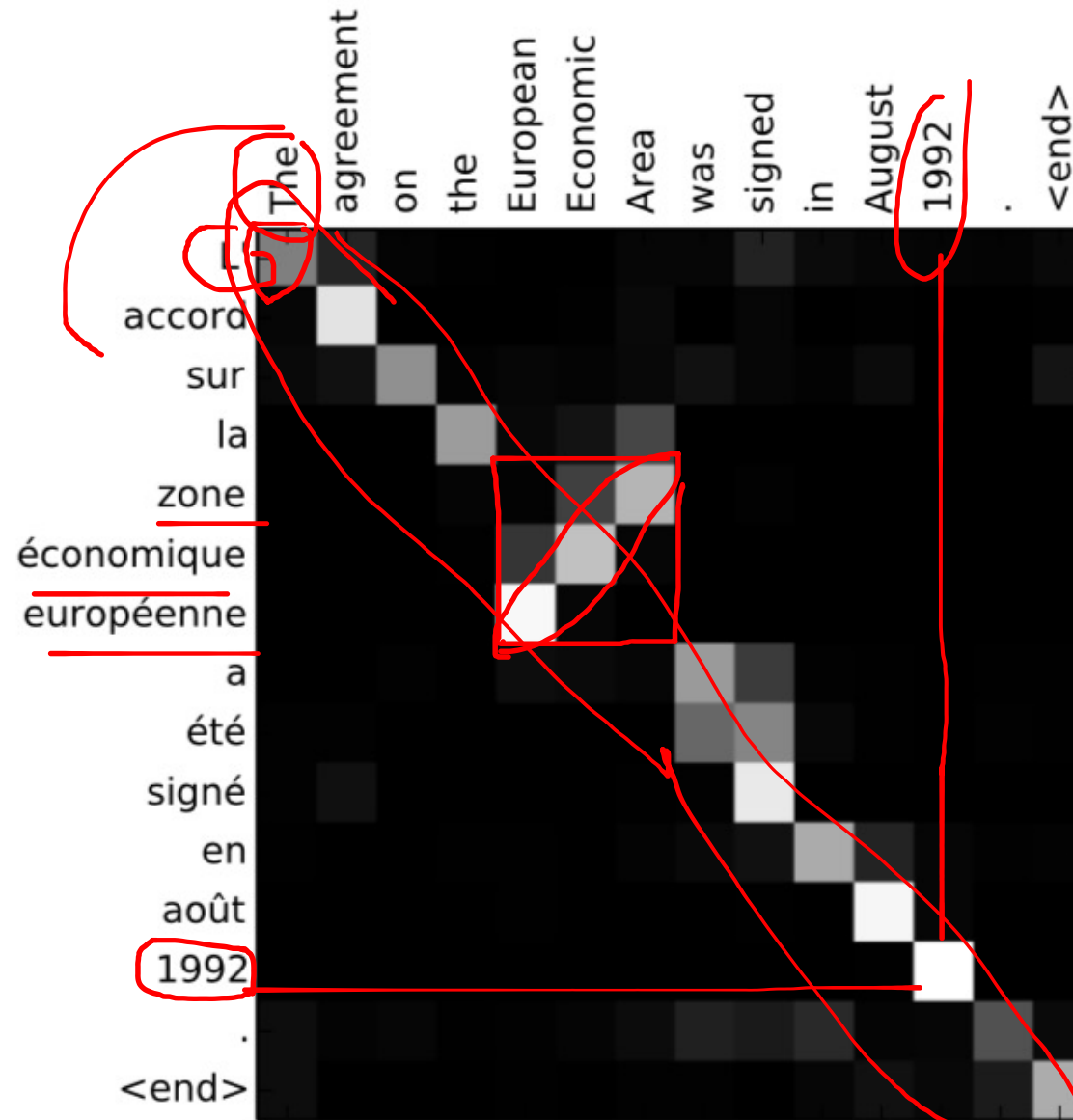
가/2

(Bidirectional) encoder

$$a^{<t'>} = (\vec{a}^{<t'>}, \tilde{a}^{<t'>})$$



Attention



References

- Sequence to sequence models, Alireza Akhavan Pour
- https://d2l.ai/chapter_recurrent-modern/beam-search.html

