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Sequence to
sequence models

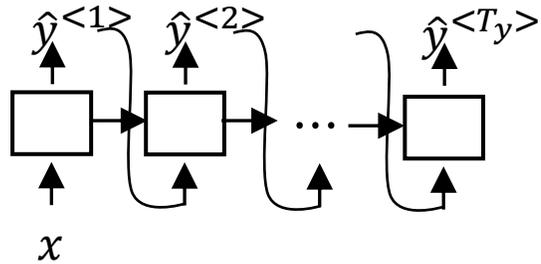
Transformers
Intuition

Transformers Motivation

Increased complexity,
sequential

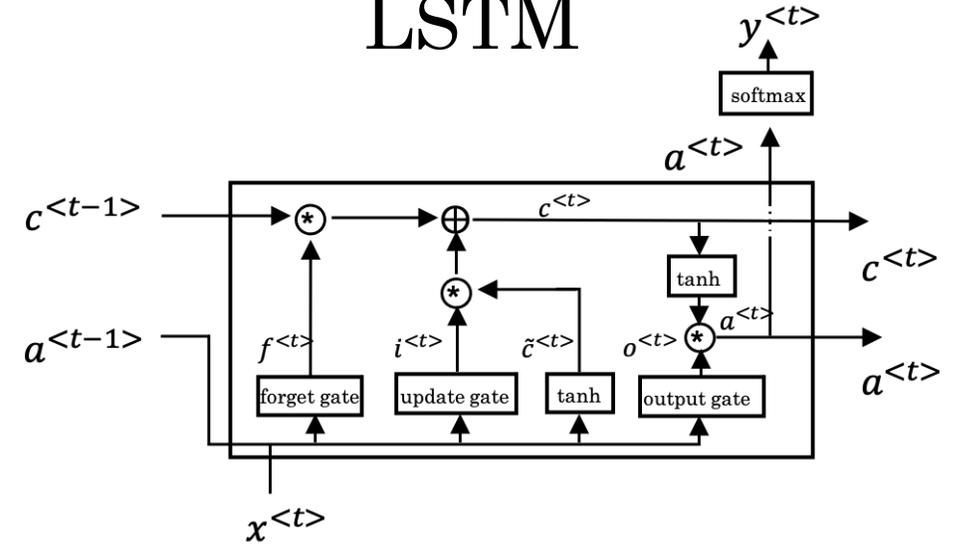


RNN



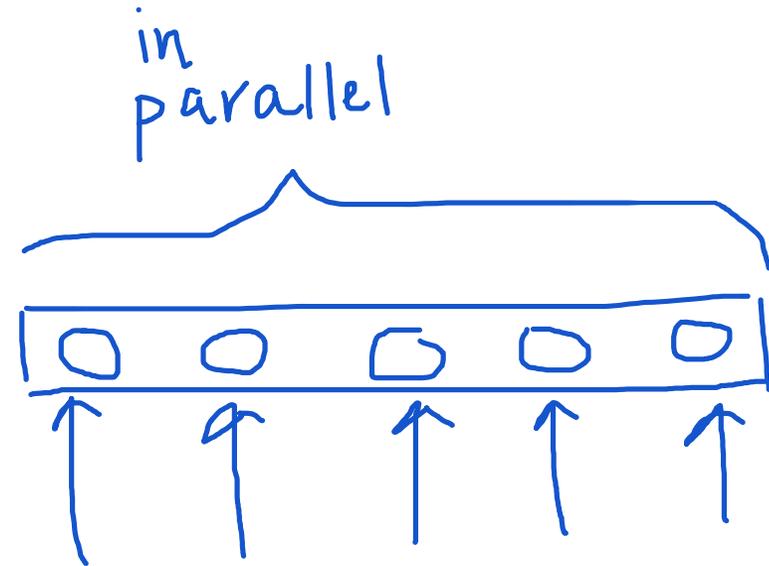
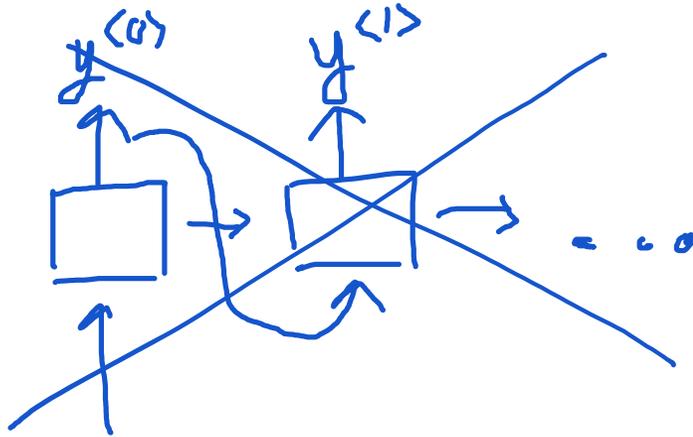
GRU

LSTM



Transformers Intuition

- Attention + CNN
 - Self-Attention
 - Multi-Head Attention





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Sequence to
sequence models

Self-Attention

Self-Attention Intuition

$A(q, K, V)$ = attention-based vector representation of a word

↪ calculate for each word

RNN Attention

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

$x^{<1>}$
Jane

$x^{<2>}$
visite

$x^{<3>}$
l'Afrique

$x^{<4>}$
en

$x^{<5>}$
septembre

Transformers Attention

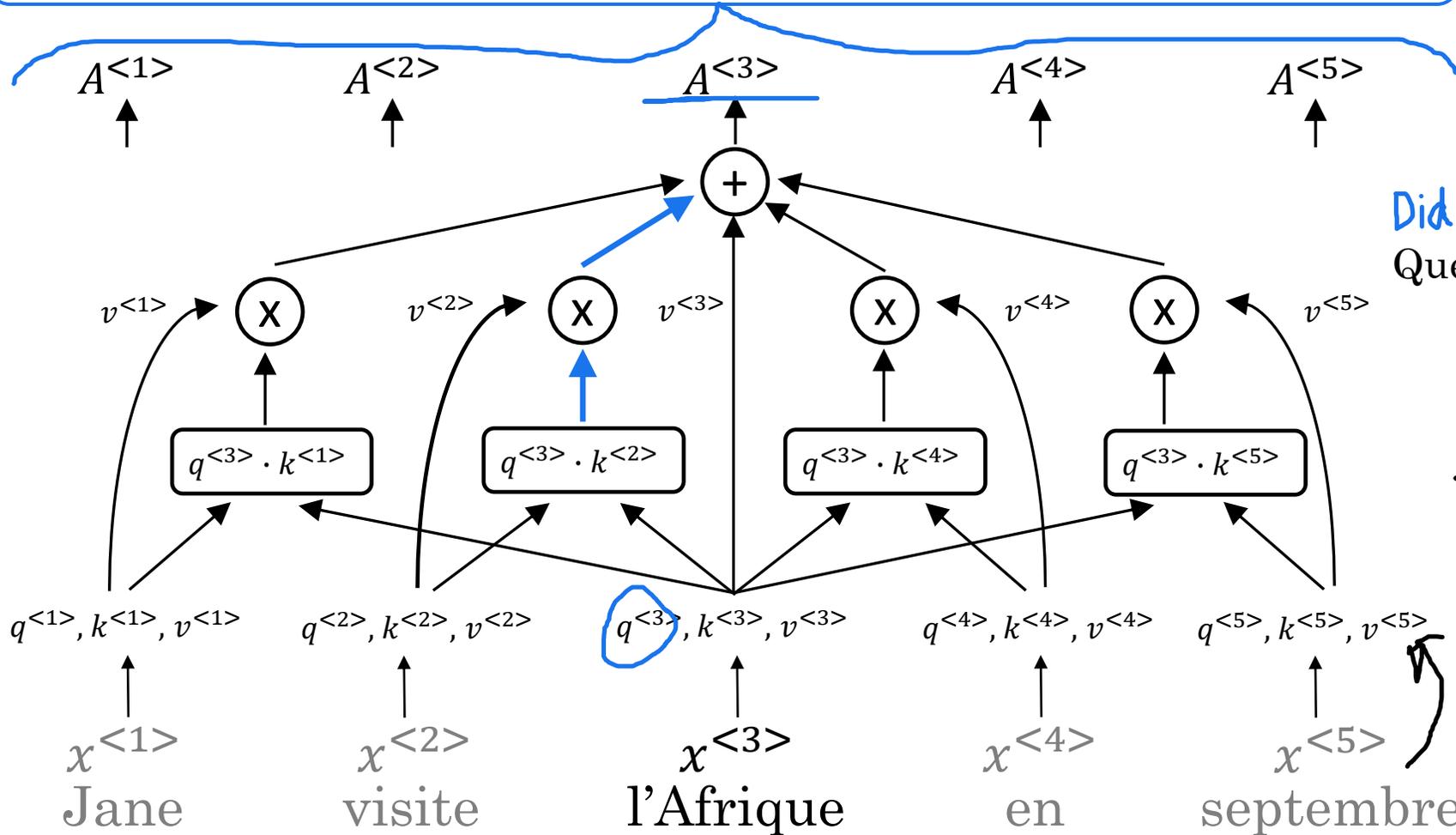
$$A(q, K, V) = \sum_i \frac{\exp(e^{<q \cdot k^{<i>}\rangle})}{\sum_j \exp(e^{<q \cdot k^{<j>}\rangle})} v^{<i>}$$

Self-Attention

$$A(q, K, V) = \sum_i \frac{\exp(e^{q \cdot k^{<i>}})}{\sum_j \exp(e^{q \cdot k^{<j>}})} v^{<i>}$$

softmax

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



Did what?

| Query (Q) | Key (K) | Value (V) |
|-----------|-----------|-----------|
| $q^{<1>}$ | $k^{<1>}$ | $v^{<1>}$ |
| $q^{<2>}$ | $k^{<2>}$ | $v^{<2>}$ |
| $q^{<3>}$ | $k^{<3>}$ | $v^{<3>}$ |
| $q^{<4>}$ | $k^{<4>}$ | $v^{<4>}$ |
| $q^{<5>}$ | $k^{<5>}$ | $v^{<5>}$ |

Handwritten notes:
 - Next to $k^{<2>}$: *person*
 - Next to $k^{<3>}$: *action*
 - Next to $v^{<1>}$: *Jane*
 - Next to $v^{<2>}$: *visit*
 - Next to $q^{<3>}$: *What's happening there?*

W^Q, W^K, W^V

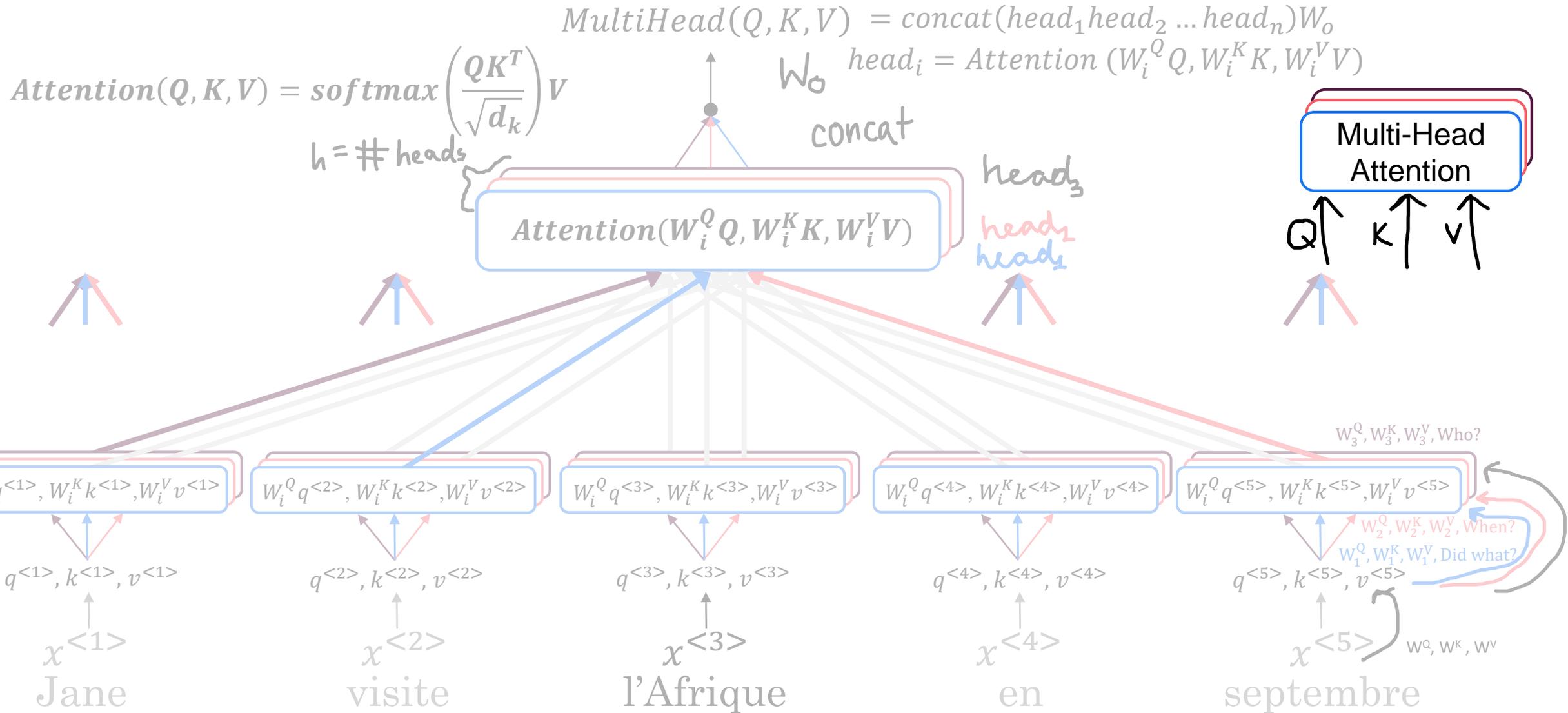


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Sequence to
sequence models

**Multi-Head
Attention**

Multi-Head Attention





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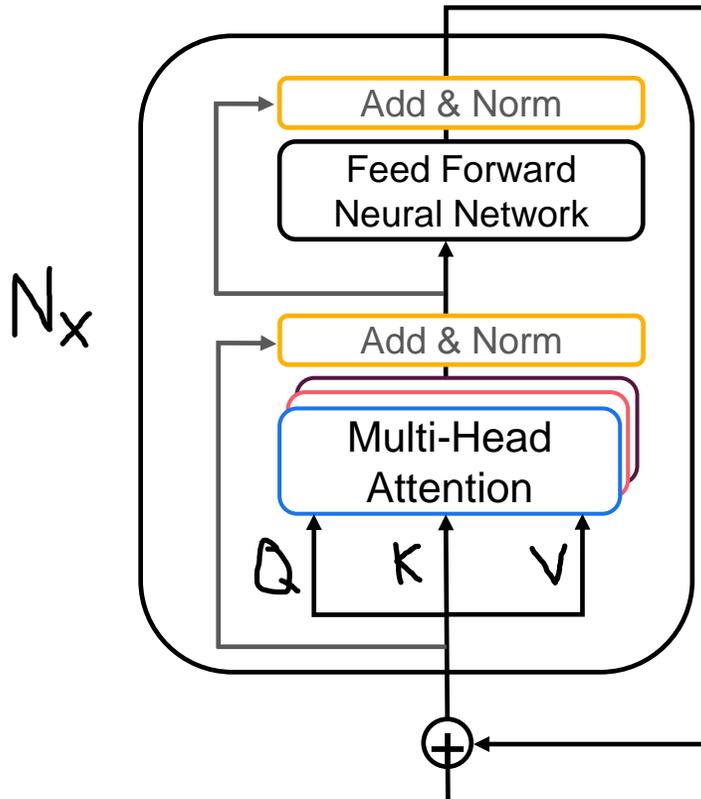
Sequence to
sequence models

Transformers

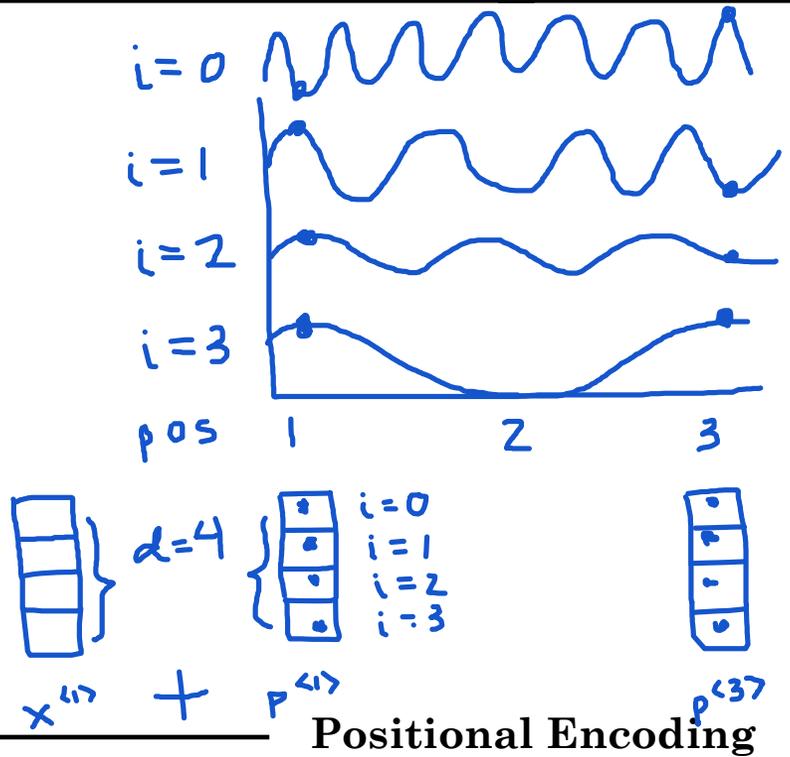
Transformer Details

<SOS> Jane visits Africa in September <EOS>

Encoder



<SOS> $x^{<1>}$ $x^{<2>}$... $x^{<T_x-1>}$ $x^{<T_x>}$ <EOS>
Jane visite l'Afrique en septembre

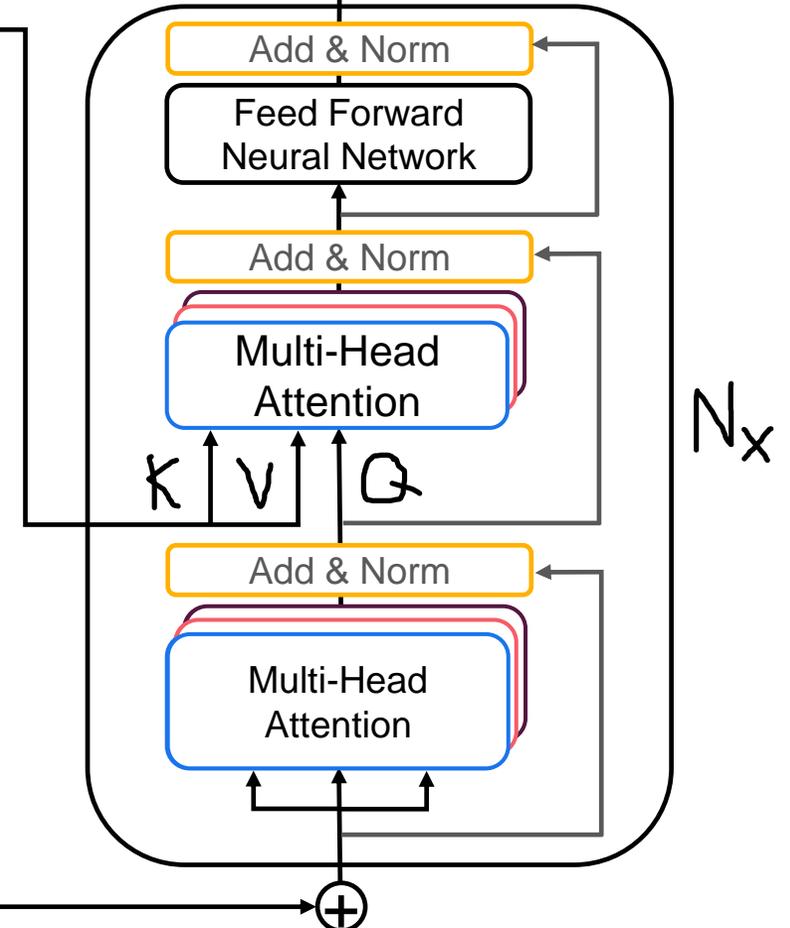


Positional Encoding

$$PE_{(pos,2i)} = \sin\left(\frac{pos}{1000^{\frac{2i}{d}}}\right)$$

$$PE_{(pos,2i+1)} = \cos\left(\frac{pos}{1000^{\frac{2i}{d}}}\right)$$

Decoder



<SOS> $y^{<1>}$ $y^{<2>}$... $y^{<T_y-1>}$ $y^{<T_y>}$
<SOS> Jane visits Africa in September

