

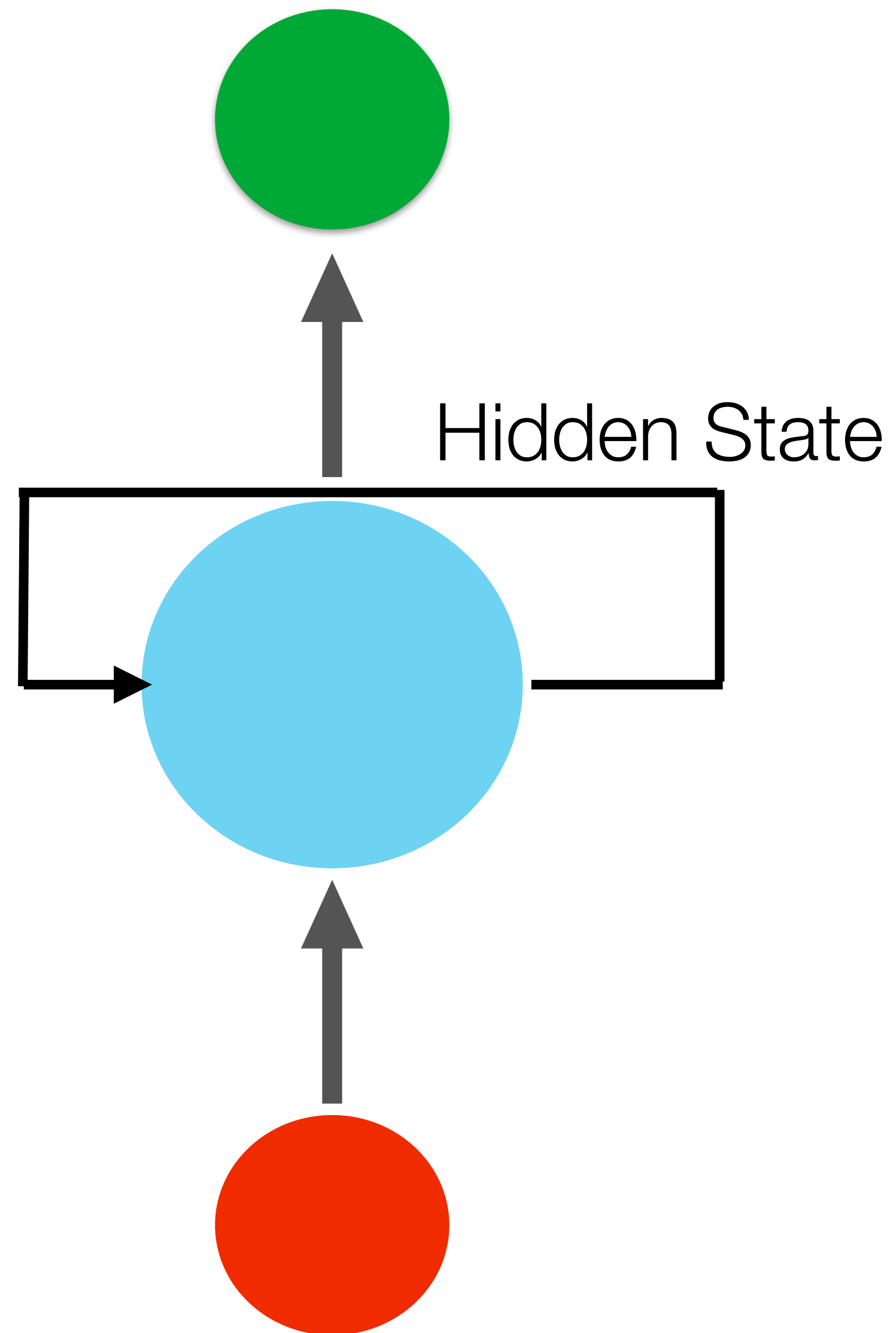


Attention Is All You Need

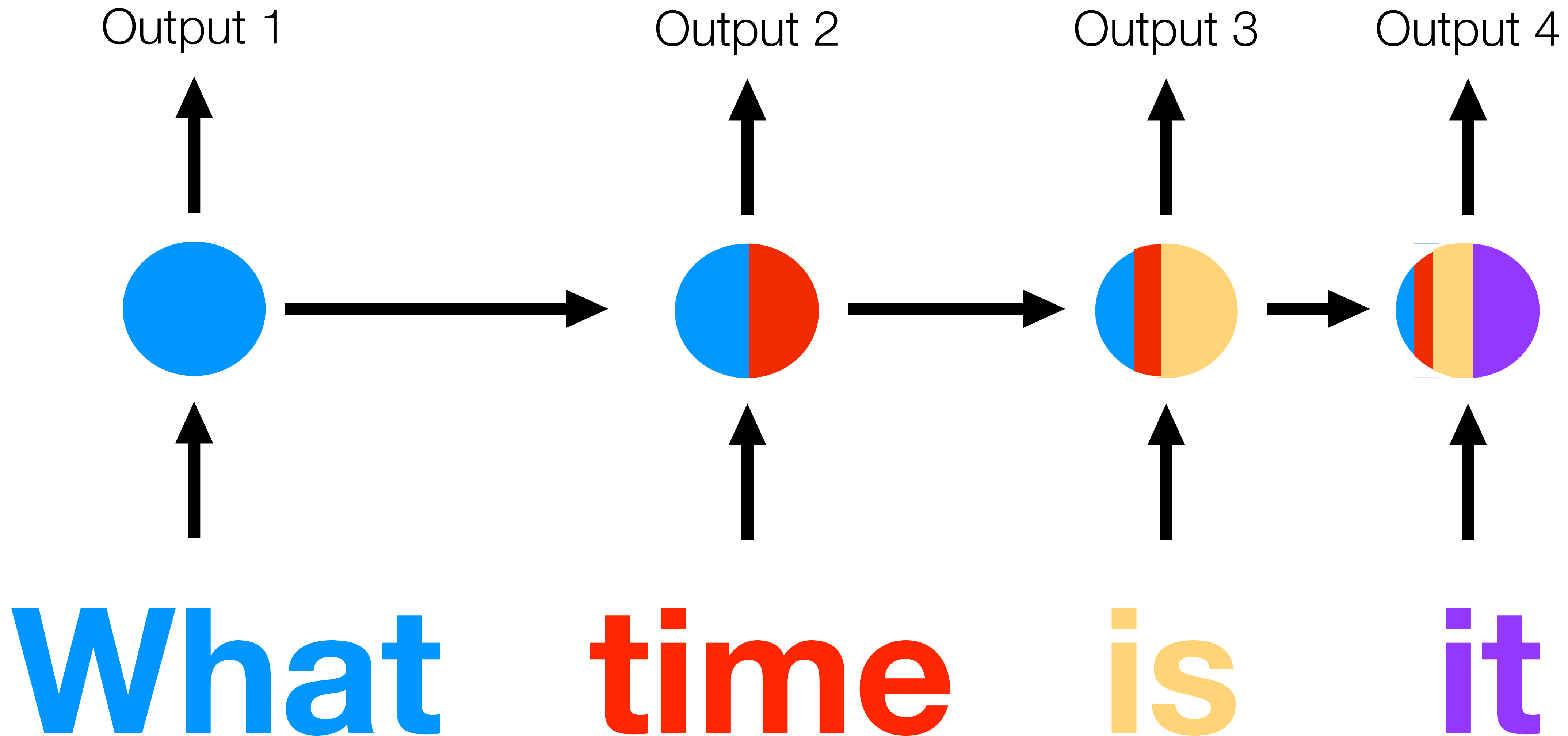
Matteo Omenetti

Google

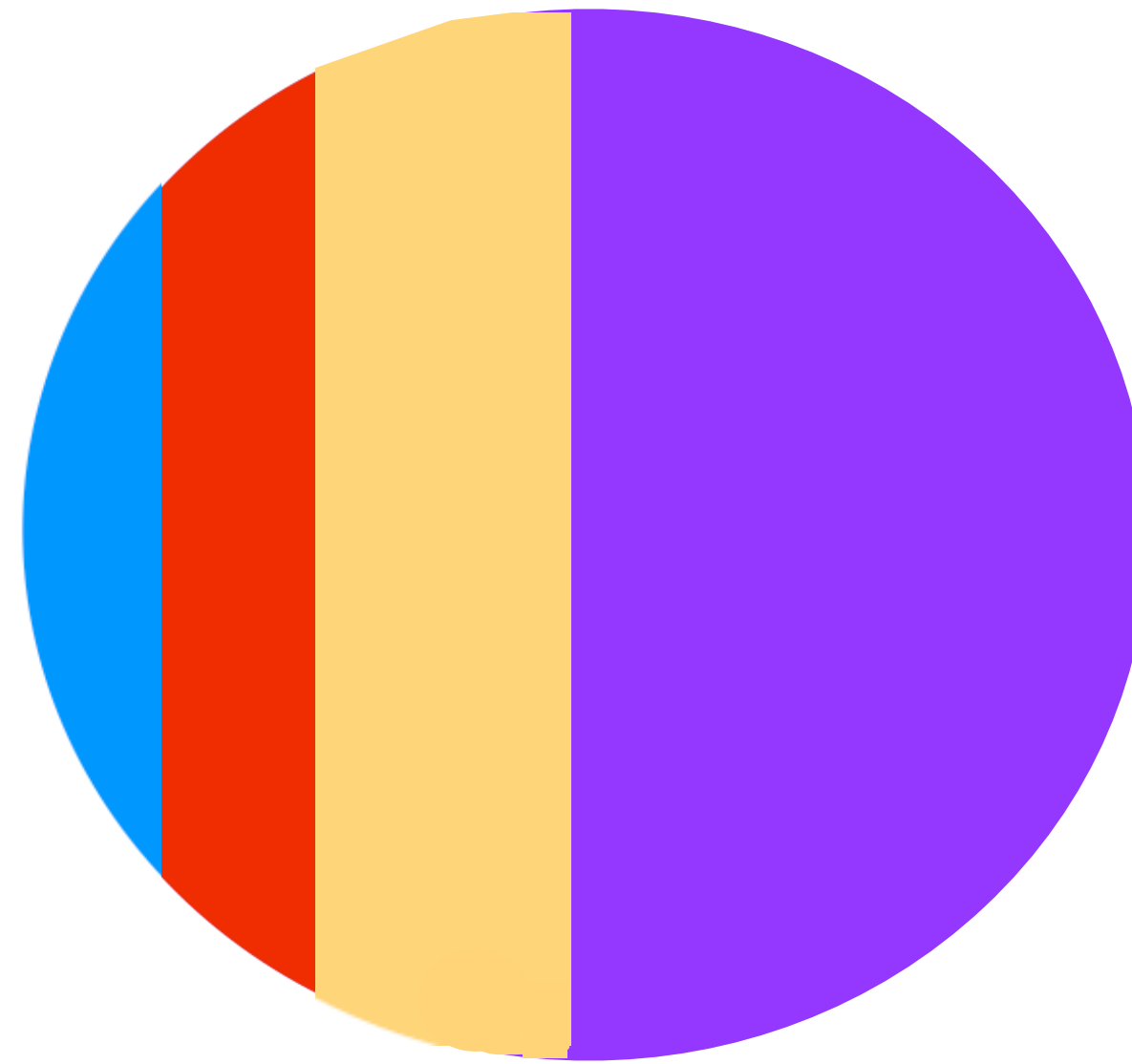
Recurrent Neural Networks (RNNs) (1)



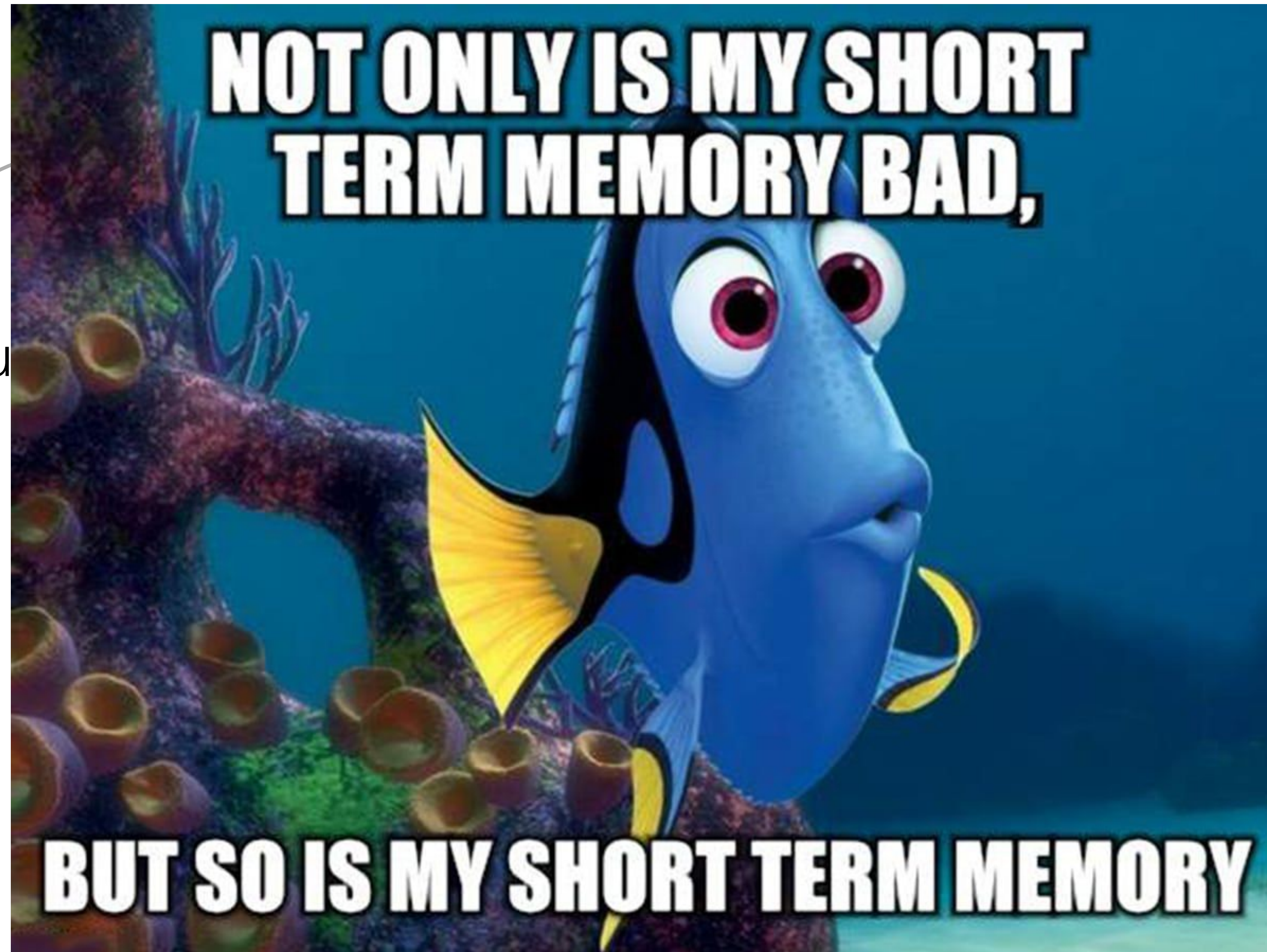
Recurrent Neural Networks (RNNs) (2)



Recurrent Neural Networks (RNNs) (3)



Recurrent Neural Networks (RNNs) (3)

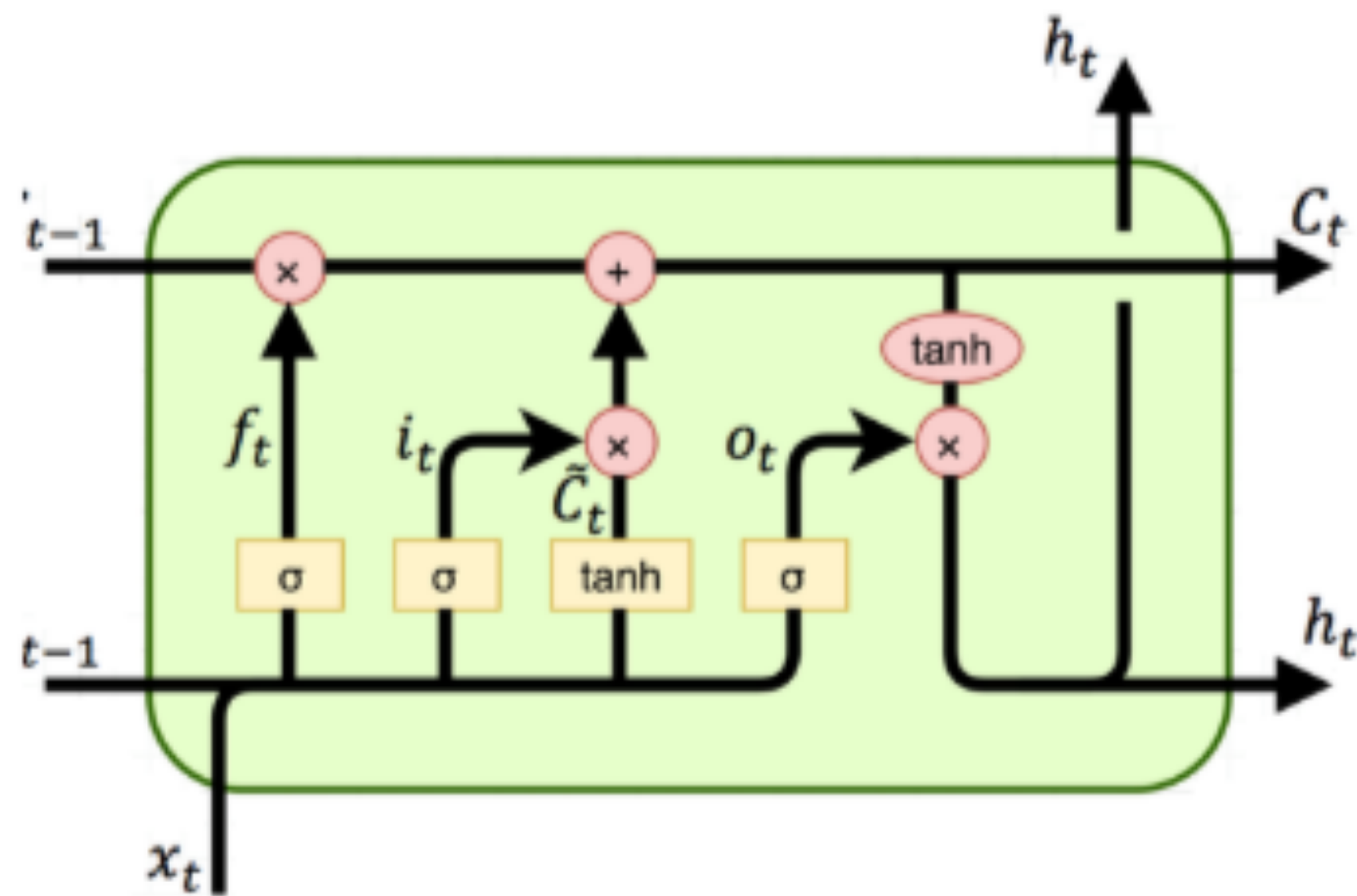


As aliens entered our

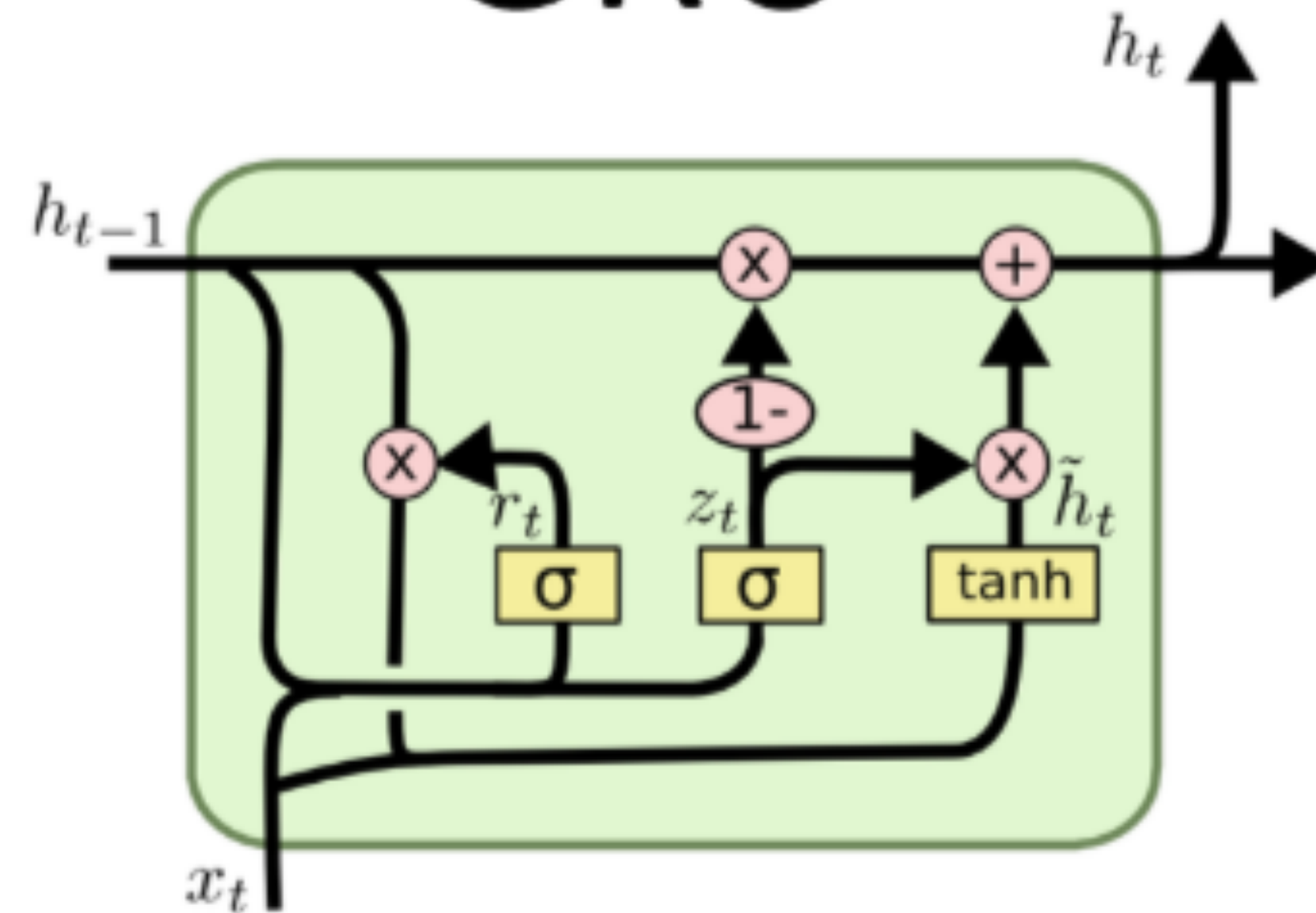
of extraterrestrials...

Recurrent Neural Networks (RNNs) (4)

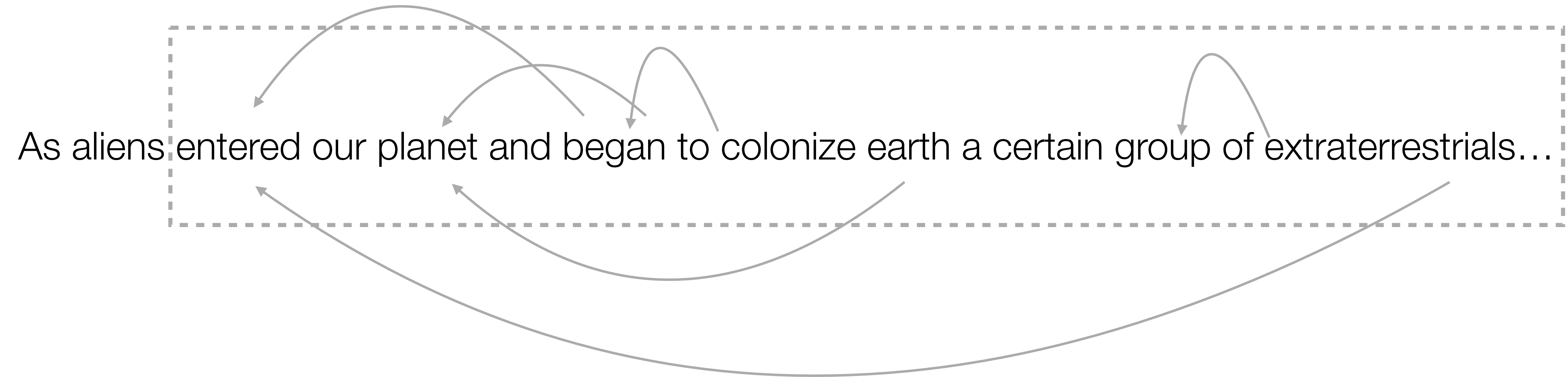
LSTM



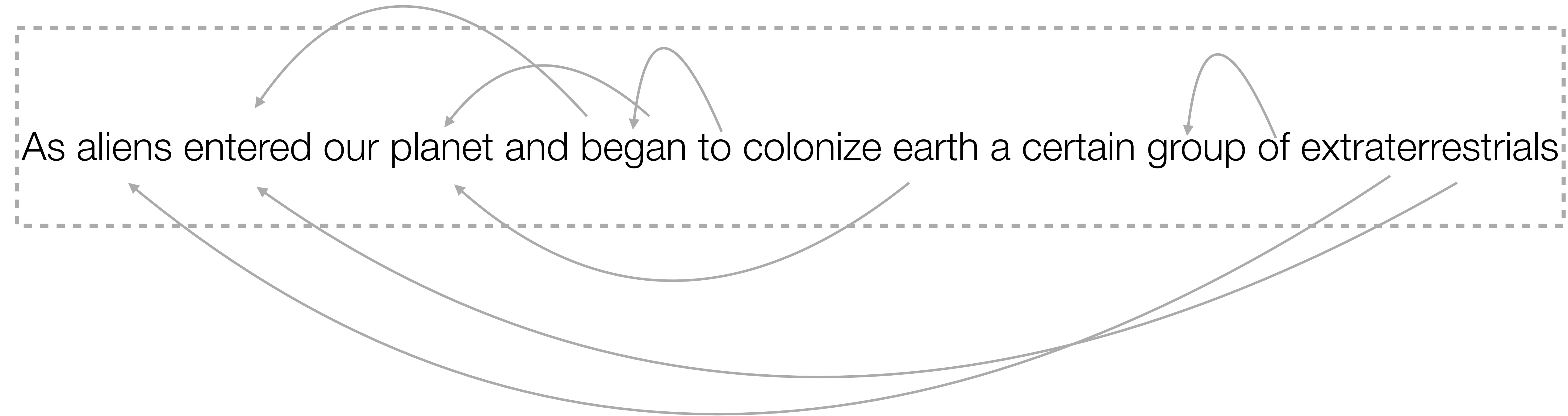
GRU



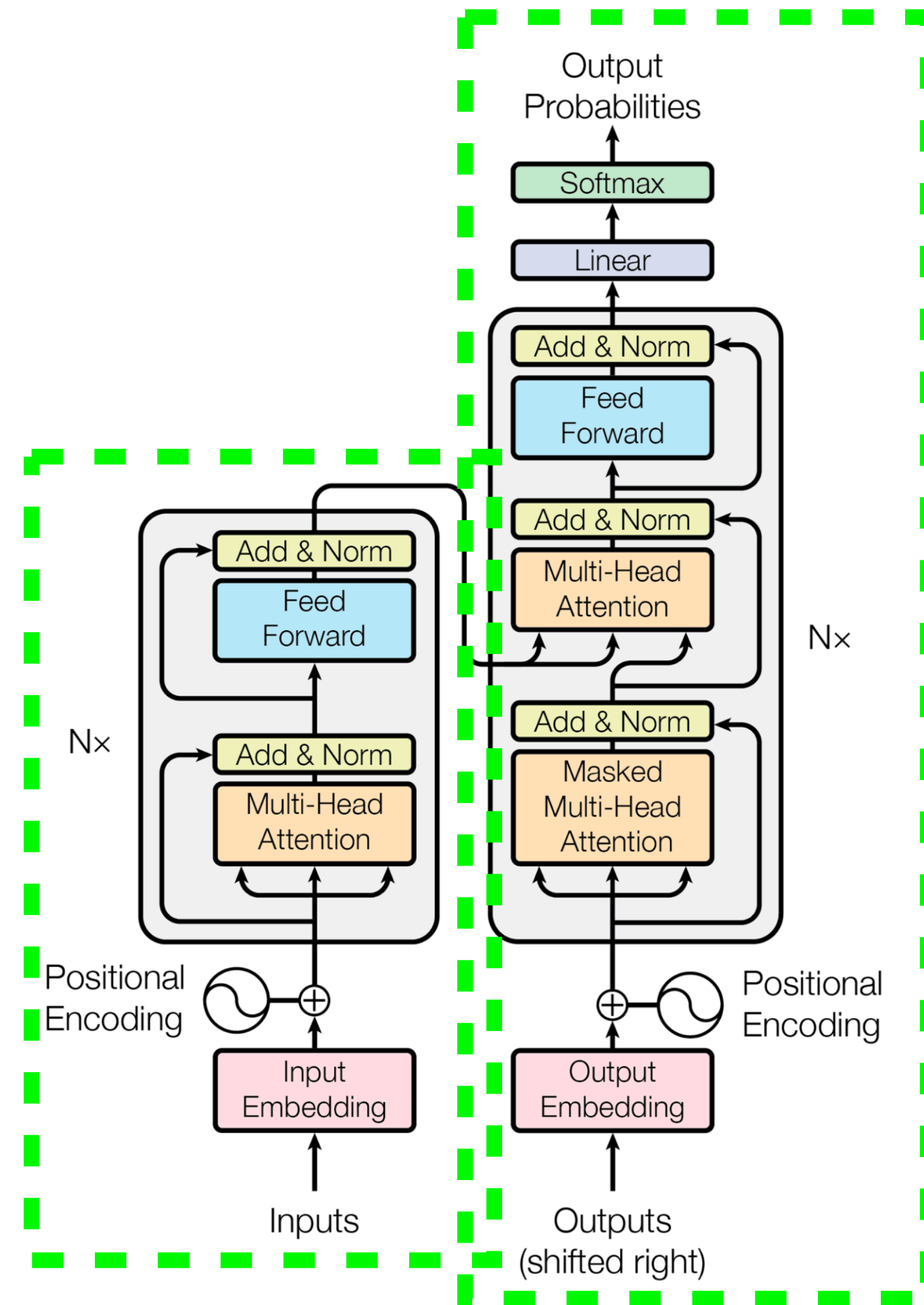
Recurrent Neural Networks (RNNs) (5)



Transformers




The Architecture



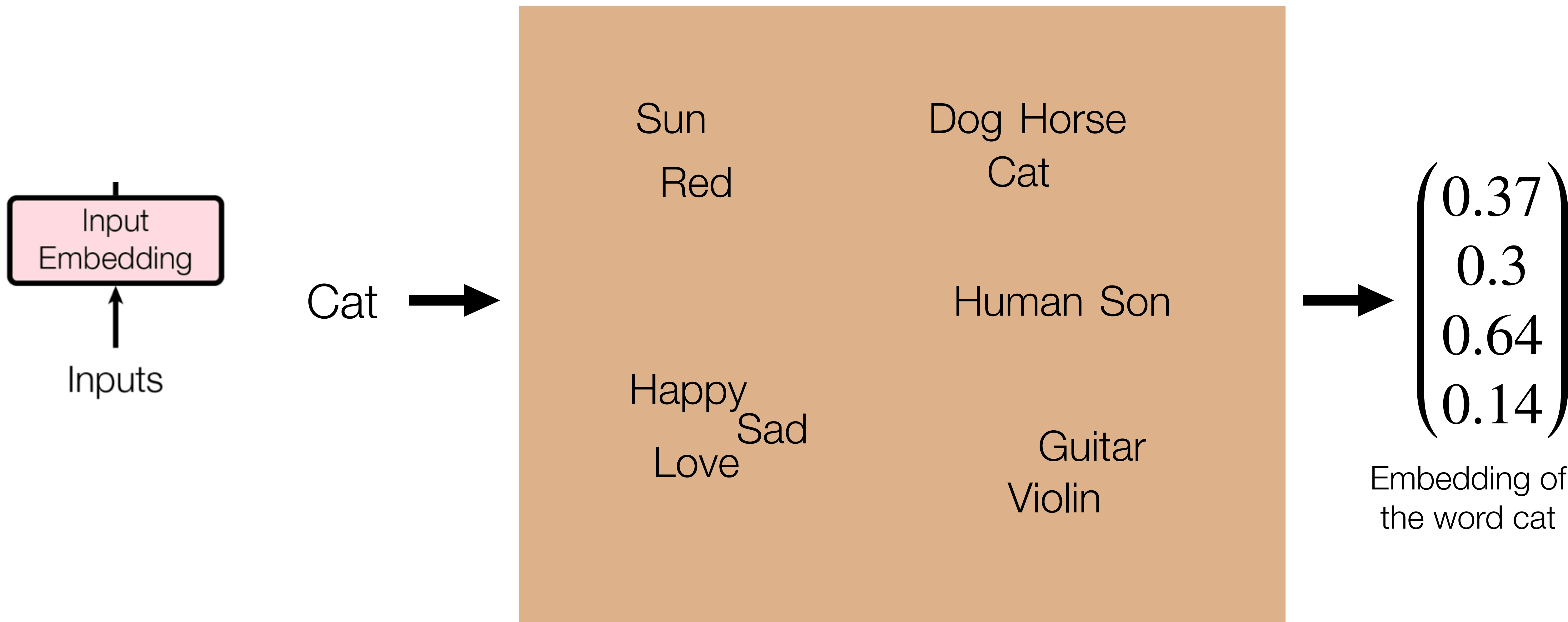
The Example

 The cat is under the table

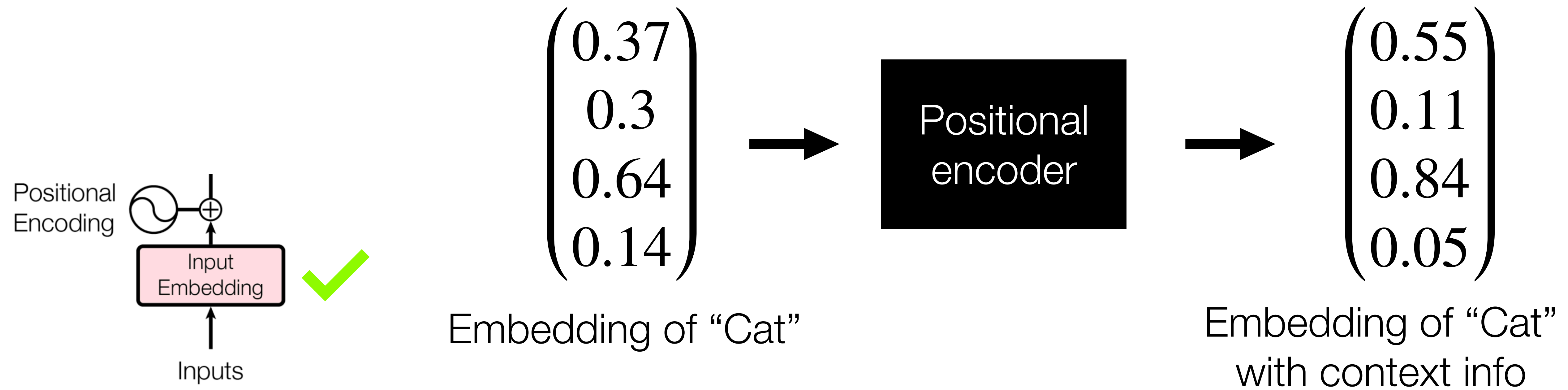
=

 Il gatto è sotto il tavolo

Input Embedding

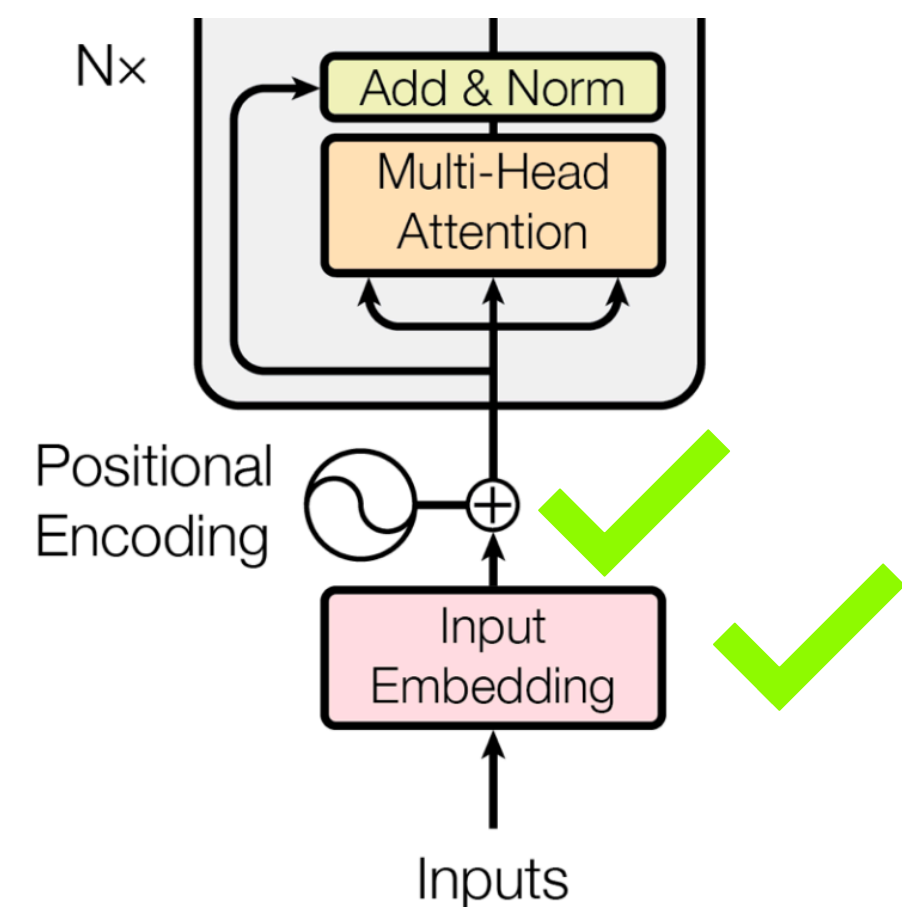


Positional Encoding



$$PE_{(pos, 2i)} = \sin(pos/10000^{2i/d_{\text{model}}})$$
$$PE_{(pos, 2i+1)} = \cos(pos/10000^{2i/d_{\text{model}}})$$

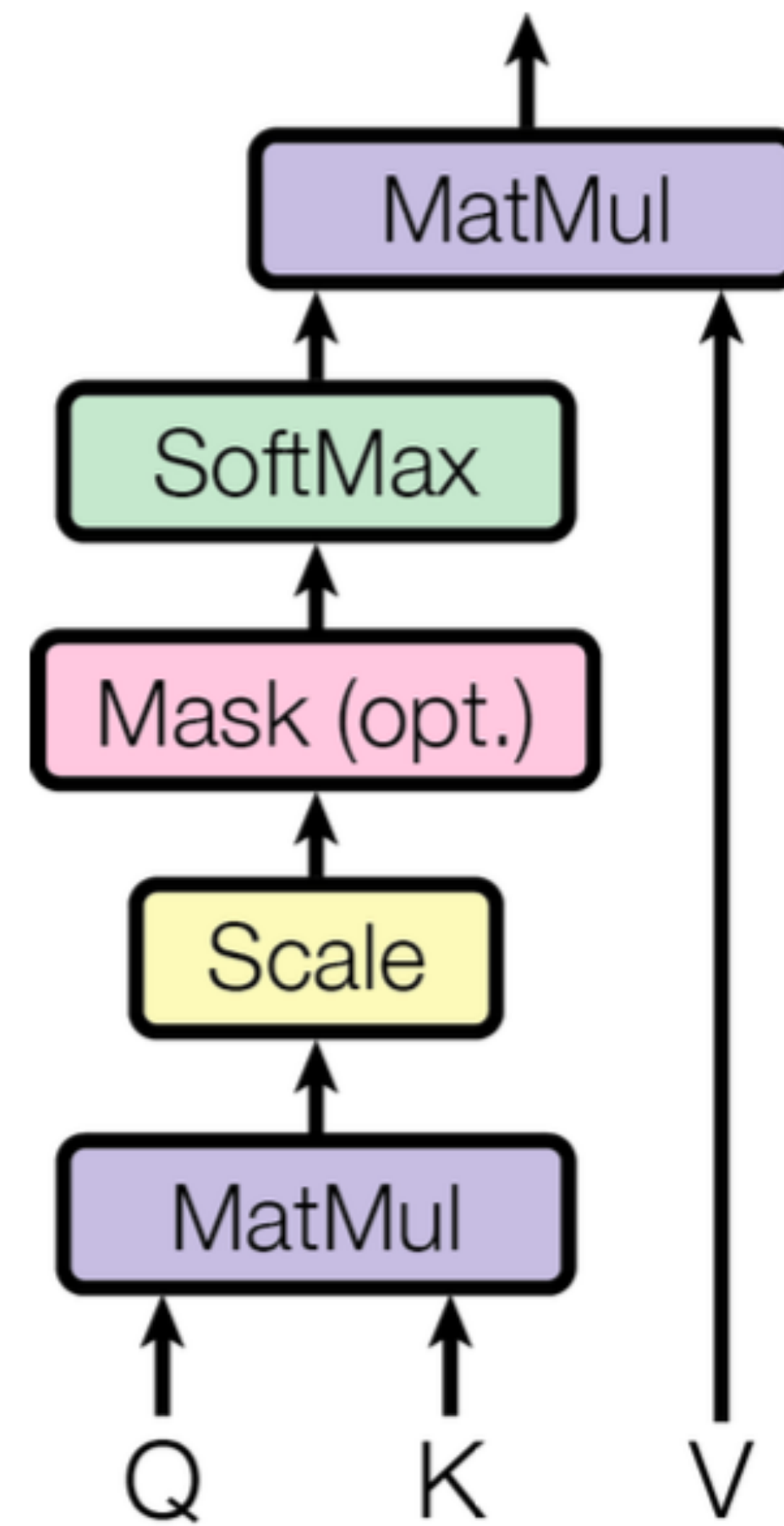
Multi-Head Attention (1)



Attention Vectors

The	→	The	cat	is	under	the	table	[0.70 0.10 0.05 0.09 0.2 0.04]
cat	→	The	cat	is	under	the	table	[0.13 0.64 0.02 0.10 0.01 0.10]
is	→	The	cat	is	under	the	table	▪
under	→	The	cat	is	under	the	table	▪
the	→	The	cat	is	under	the	table	▪
table	→	The	cat	is	under	the	table	

Multi-Head Attention (4)



Multi-Head Attention (3)

T	c	i	u	t	t
h	a	s	n	h	a
e	t		d	e	b
			e		l
			r		e

Query matrix

X

The
cat
is
under
the
table

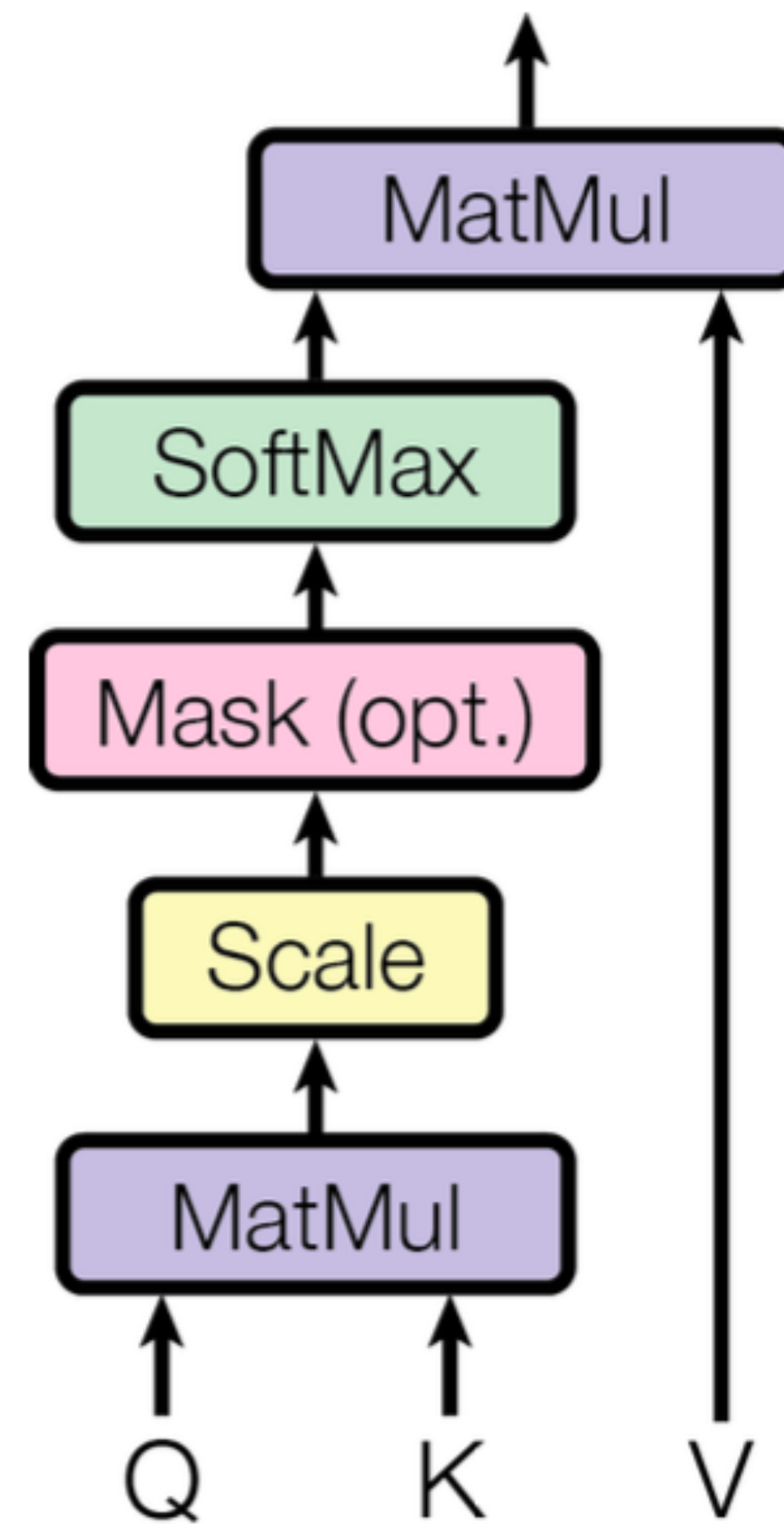
Key matrix

=

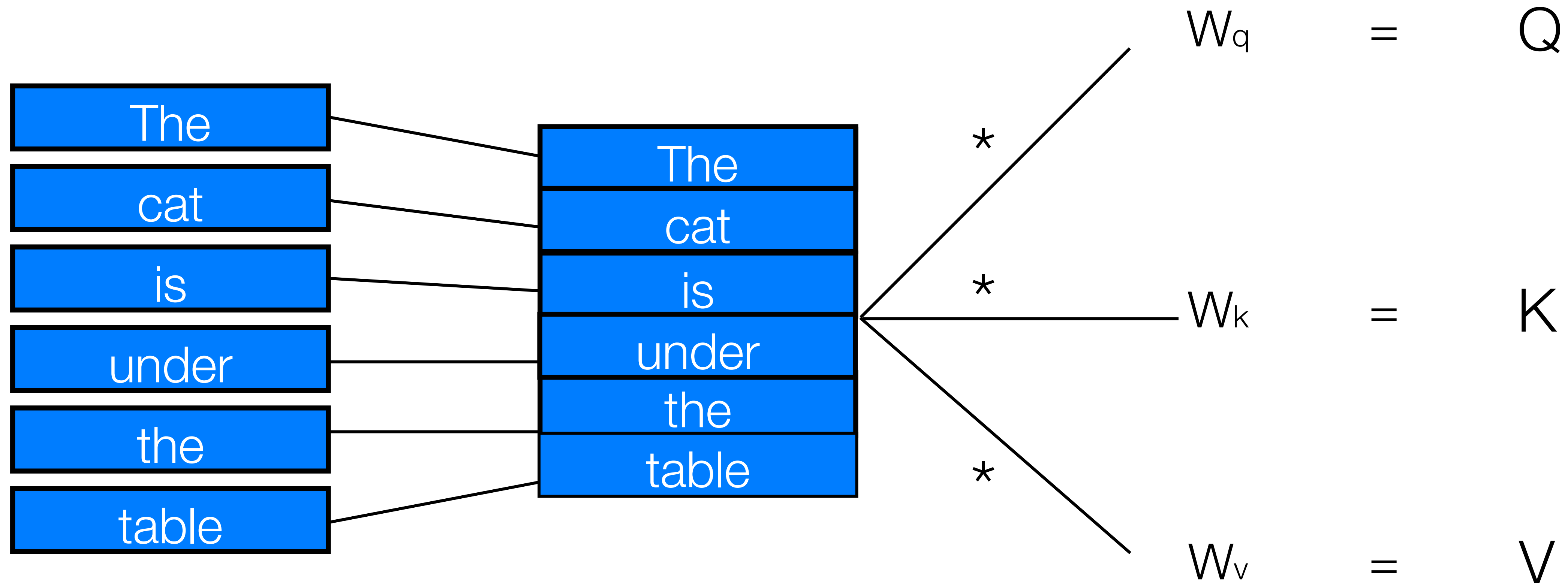
	The	cat	is	under	the	table
The	98	27	10	12	5	11
cat	27	89	31	67	5	15
is	6	20	77	10	6	10
under	5	20	5	74	8	30
the	5	9	3	20	80	30
table	5	20	8	25	20	80

Score matrix

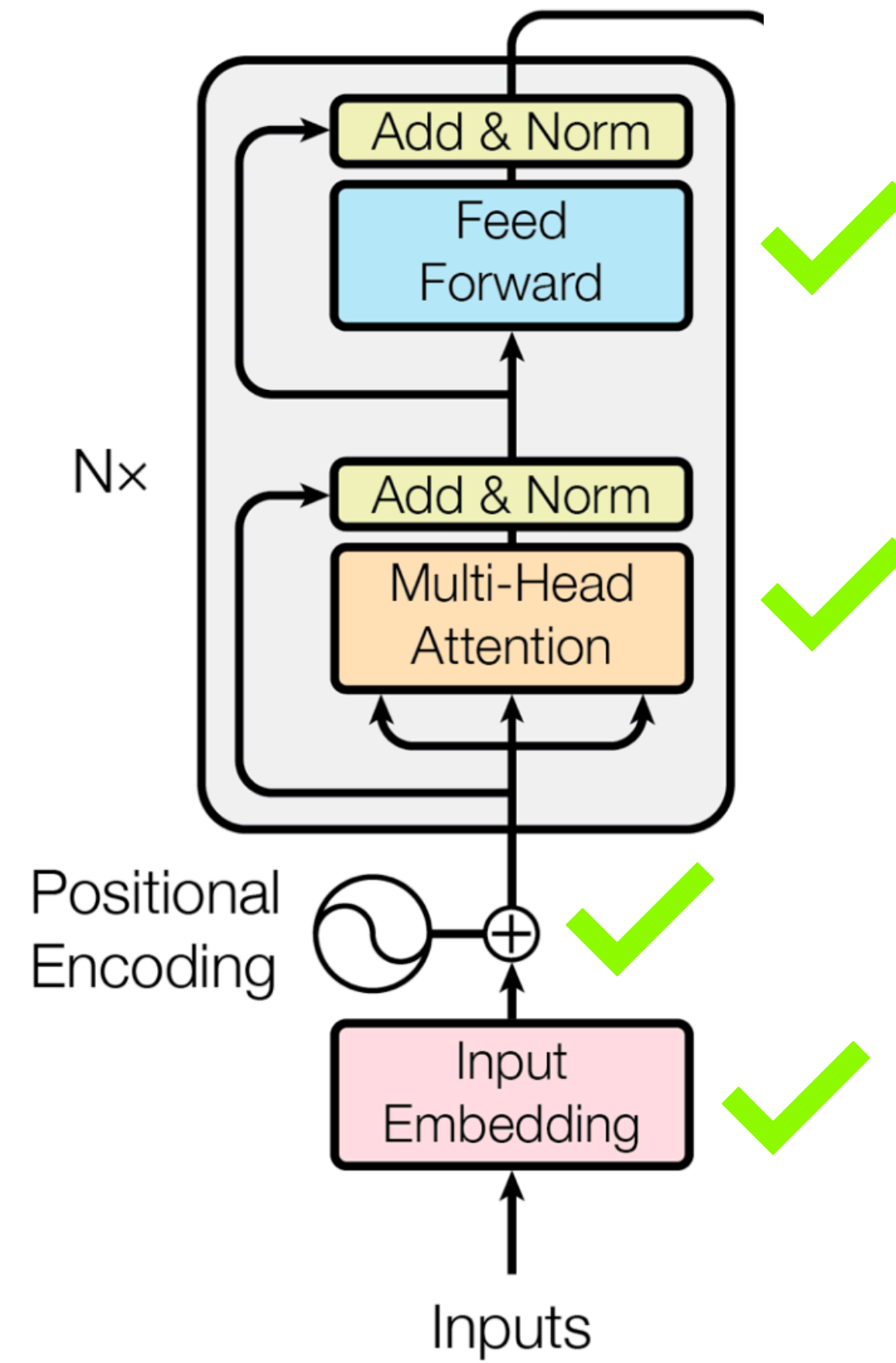
Multi-Head Attention (4)



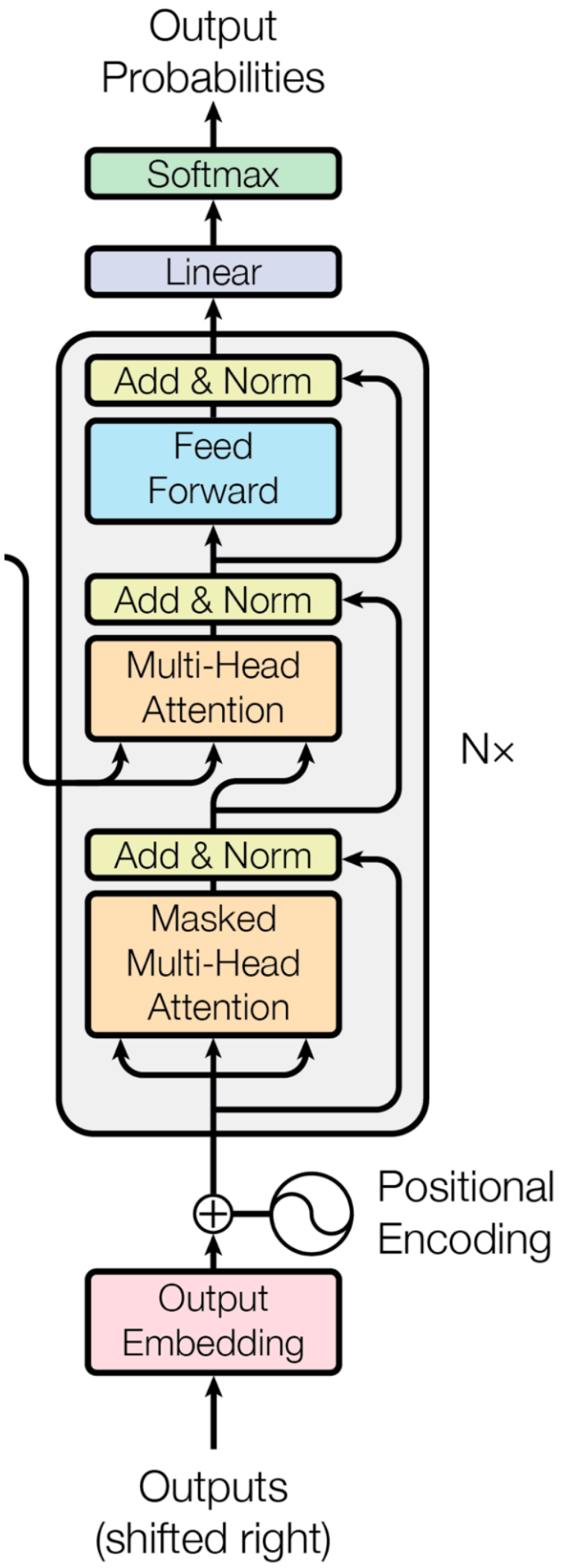
Multi-Head Attention (5)



Feed Forward



Decoder



Masked Multi-Head Attention

	Il	gatto	è	sotto	il	tavolo
Il	98					
gatto	27	89				
è	6	20	77			
sotto	5	20	5	74		
il	5	9	3	20	80	
tavolo	5	20	8	25	20	80

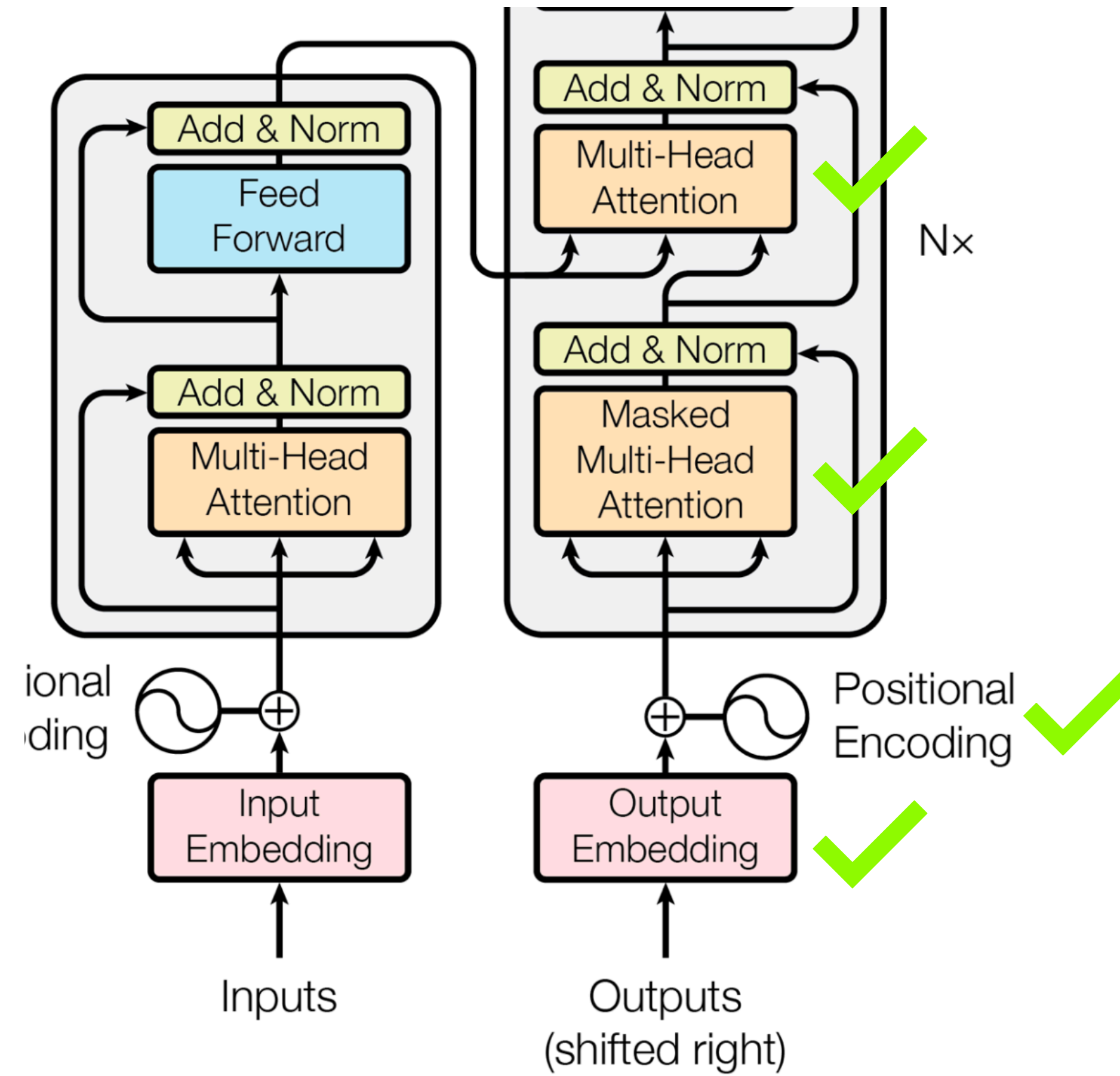
Score matrix

+

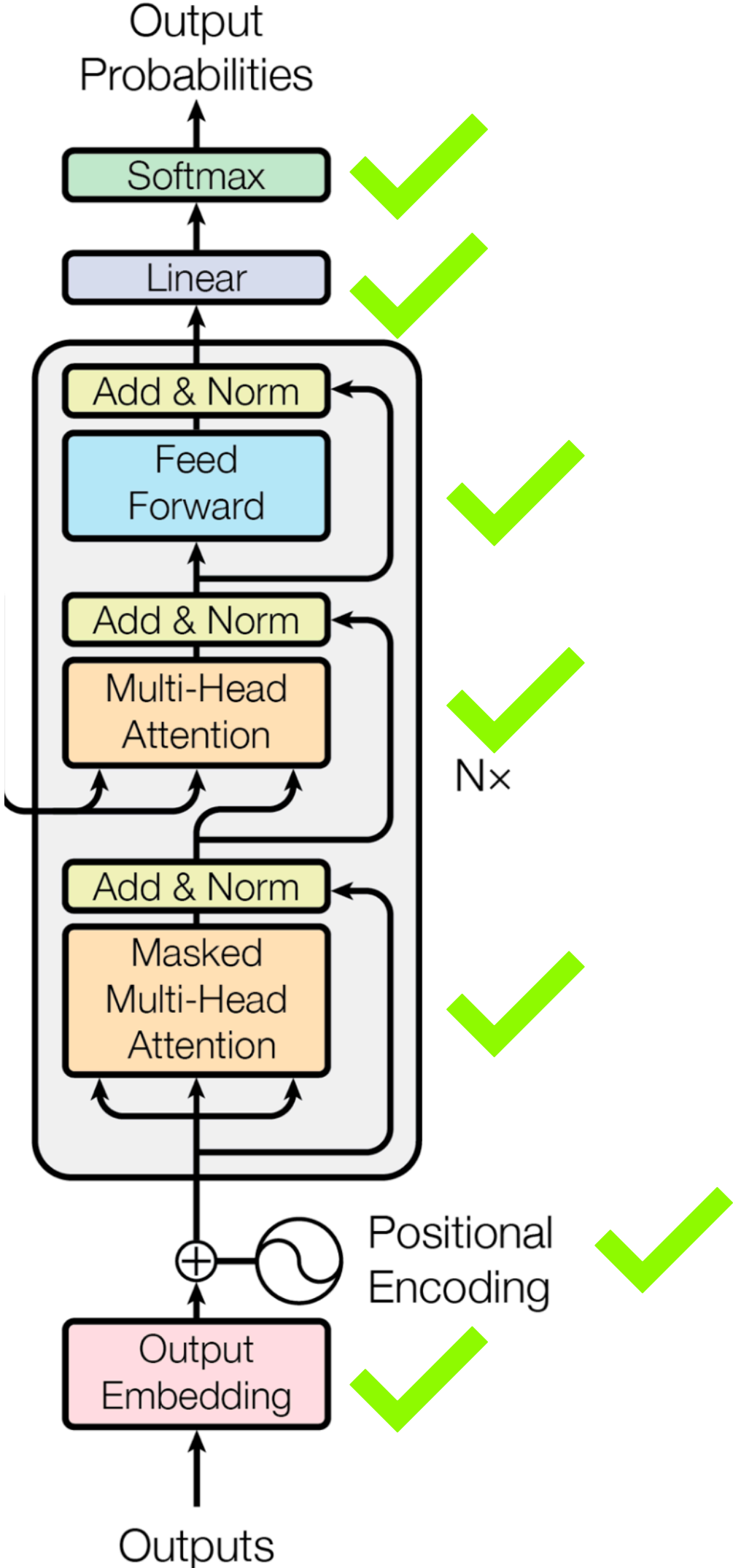
0	-inf	-inf	-inf	-inf	-inf
0	0	-inf	-inf	-inf	-inf
0	0	0	-inf	-inf	-inf
0	0	0	0	-inf	-inf
0	0	0	0	0	-inf
0	0	0	0	0	0

Mask matrix

Multi-Head Attention



Multi-Head Attention



Results

Model	BLEU		Training Cost (FLOPs)	
	EN-DE	EN-FR	EN-DE	EN-FR
ByteNet [18]	23.75			
Deep-Att + PosUnk [39]		39.2		$1.0 \cdot 10^{20}$
GNMT + RL [38]	24.6	39.92	$2.3 \cdot 10^{19}$	$1.4 \cdot 10^{20}$
ConvS2S [9]	25.16	40.46	$9.6 \cdot 10^{18}$	$1.5 \cdot 10^{20}$
MoE [32]	26.03	40.56	$2.0 \cdot 10^{19}$	$1.2 \cdot 10^{20}$
Deep-Att + PosUnk Ensemble [39]		40.4		$8.0 \cdot 10^{20}$
GNMT + RL Ensemble [38]	26.30	41.16	$1.8 \cdot 10^{20}$	$1.1 \cdot 10^{21}$
ConvS2S Ensemble [9]	26.36	41.29	$7.7 \cdot 10^{19}$	$1.2 \cdot 10^{21}$
Transformer (base model)	27.3	38.1	$3.3 \cdot 10^{18}$	
Transformer (big)	28.4	41.8	$2.3 \cdot 10^{19}$	



Thanks for your attention

Q&A