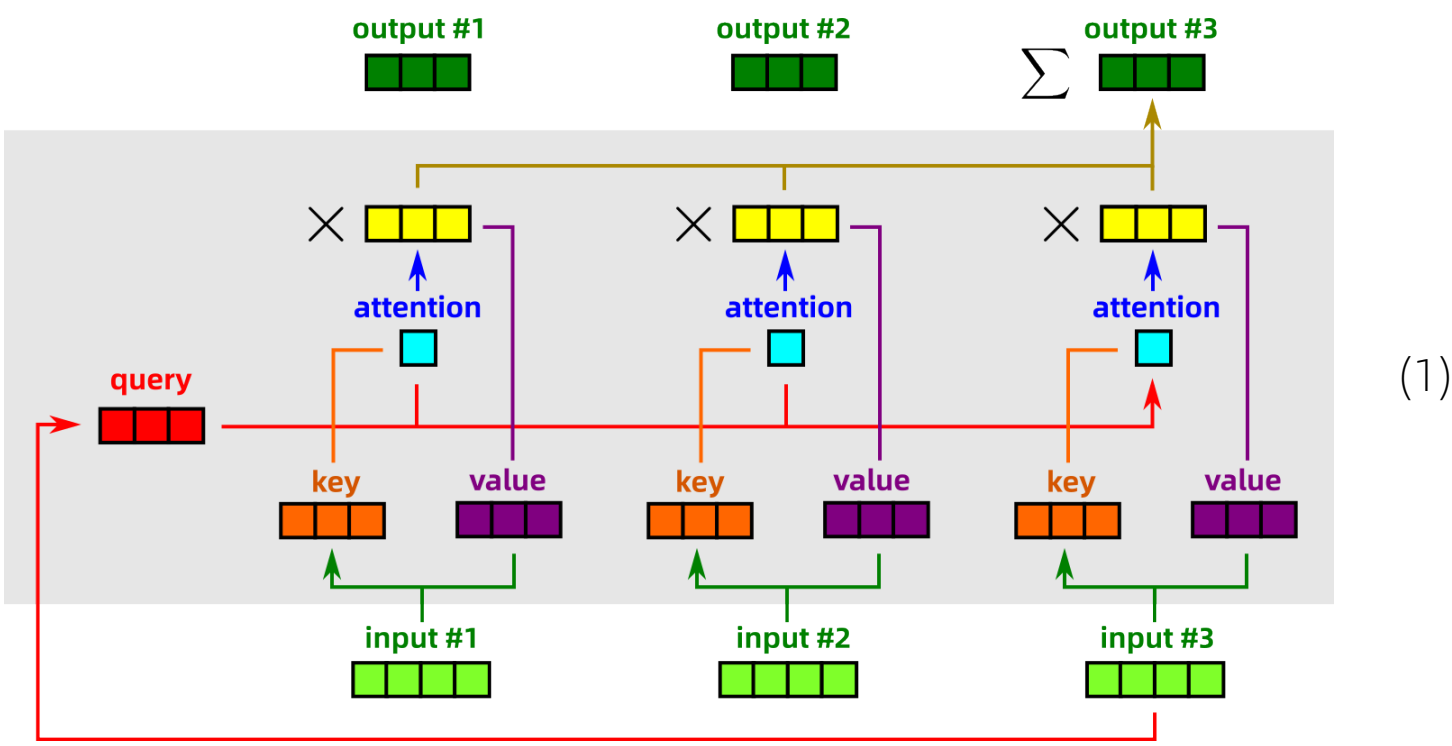


①

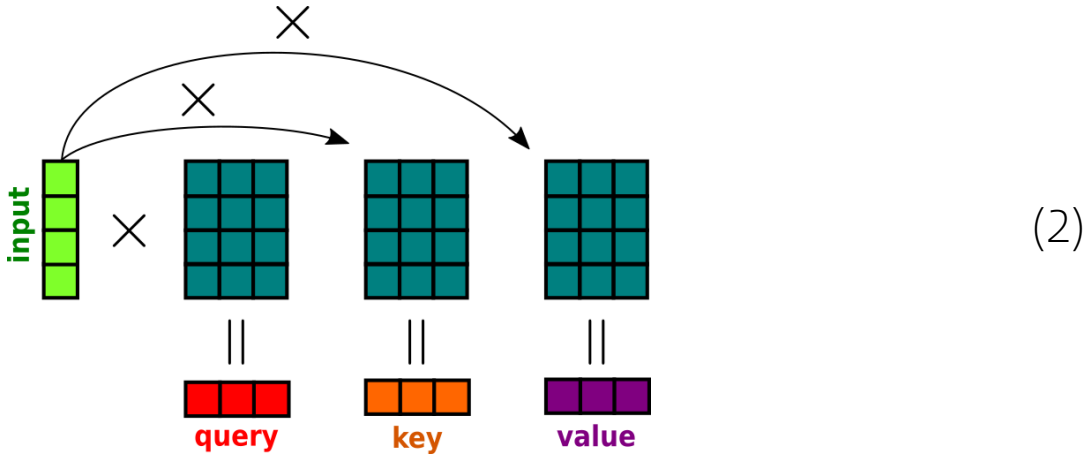
Transformer has logic structure

In this infographic I'd explain a major finding that is the culmination of many years of my research: the Transformer is a symbolic-logic machine.

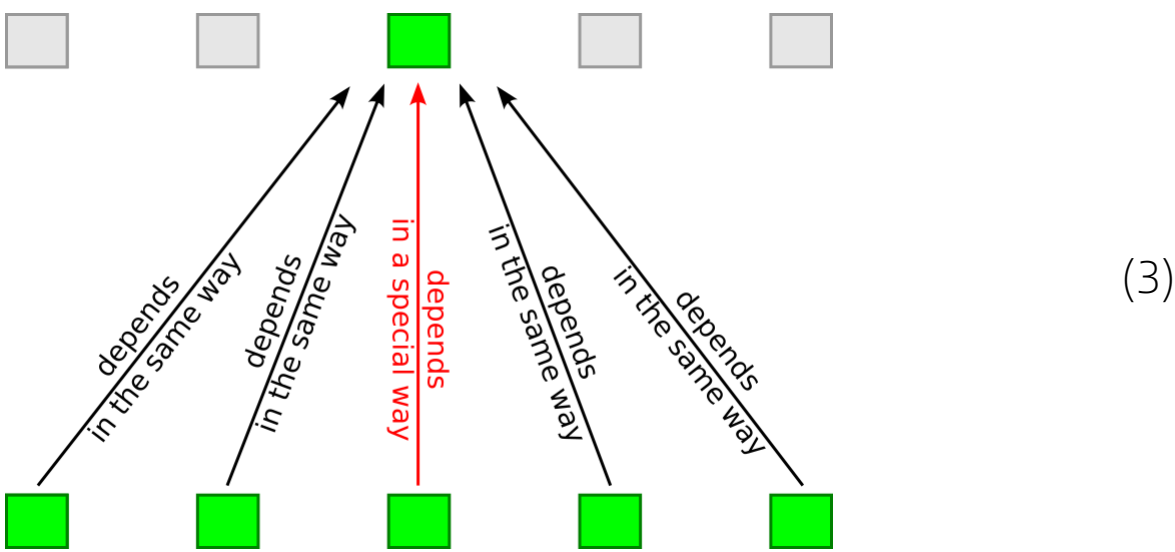
For your convenience let's refresh on the Transformer's **Self-Attention** mechanism:



“Input” tokens are translated to Q, K, V (query, key, value)’s via matrix multiplication, which can be regarded as a kind of table look-up, or **memory store**:



From an abstract point of view, the Transformer has the following structure, which gives rise to its **equivariance** property (if input elements are swapped in a certain order, the output elements changes the same way):



The equivariance property corresponds to the **exchangeability** of logic propositions:

$$A \wedge B \Leftrightarrow B \wedge A$$

(4)

For example:

$$\text{it's raining} \wedge \text{I'm heart-broken} \Leftrightarrow \text{I'm heart-broken} \wedge \text{it's raining}$$

(5)

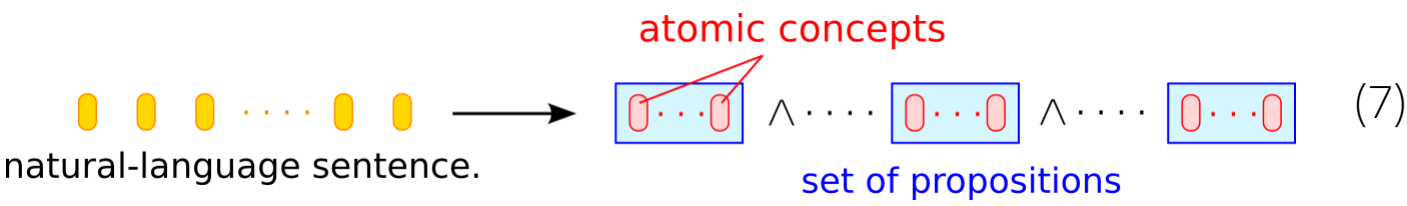
Propositions are made up of **atomic concepts**, but here, at the sub-propositional level, atoms cannot be permuted freely:

$$\text{I} \cdot \text{love} \cdot \text{you} \neq \text{you} \cdot \text{love} \cdot \text{me}$$

(6)

otherwise there would be no such things as heart-breaks.

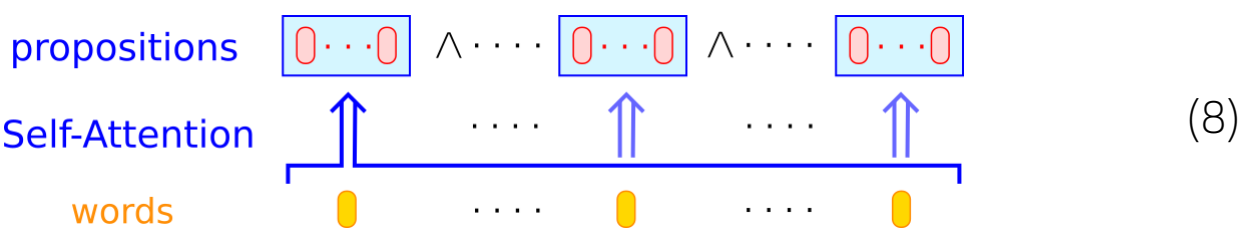
Given a natural-language sentence, we'd like to convert or **decompose** it into a bunch of logic propositions:



The structure on the right is a **mental state** of a logical AI system. It is composed of (exchangeable) propositions, which are in turn made up of atomic concepts. This 2-level structure is characteristic of all **logical** systems.

Surprisingly, I found out that the Transformer completely satisfies this 2-level logic structure.

On the first layer, a Transformer transforms each input word token into one proposition:



The crucial point here is that propositions are made up of atoms (○), which is achieved in the Transformer by **adding** vectors (that represent atomic concepts) together. Note also that the Transformer is equivariant, so we must add “positional encoding” to each word.