

The Mathematics of Text Structure

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Abstract

In previous work we gave a mathematical foundation, referred to as DisCoCat, for how words interact in a sentence in order to produce the meaning of that sentence. To do so, we exploited the perfect structural match of grammar and categories of meaning spaces.

Here, we give a mathematical foundation, referred to as DisCoCirc, for how sentences interact in texts in order to produce the meaning of that text. We revisit DisCoCat: while in the latter all meanings are states (i.e. have no input), in DisCoCirc word meanings are types of which the state can evolve, and sentences are gates within a circuit which update the meaning of words. Like in DisCoCat, word meanings can live in a variety of spaces e.g. propositional, vectorial, or cognitive. The compositional structure are string diagrams representing information flows, and an entire text yields a single string diagram in which word meanings lift to the meaning of an entire text.

While the developments in this paper are independent of a physical embodiment (cf. classical vs. quantum computing), both the compositional formalism and suggested meaning model are highly quantum-inspired, and implementation on a quantum computer would come with a range of benefits.

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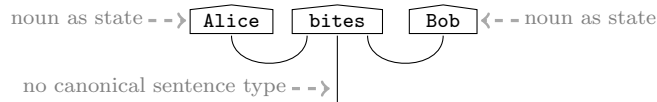
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From States to Processes

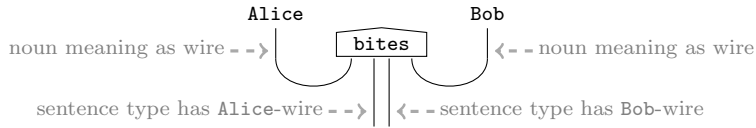
Consider the following example:

Alice is a dog.
 Bob is a person.
 Alice bites Bob.

Clearly, the meaning of the third sentence crucially depends on what we learn about the meaning of the nouns **Alice** and **Bob** in the first two sentences, turning **dog bites man** into **man bites dog** if **Bob** were to be a **dog** and **Alice** were to be a **person**. Also, before the 1st sentence is stated, **Alice** is just a meaningless name, and the same goes for **Bob** until the 2nd sentence is stated. So the meaning of **Alice** and **Bob** evolves as the text progresses, and it are the sentences that update our knowledge about **Alice** and **Bob**. What we propose is that the 3rd sentence, which would look like:



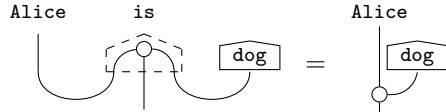
in DisCoCat, would instead be drawn like this:



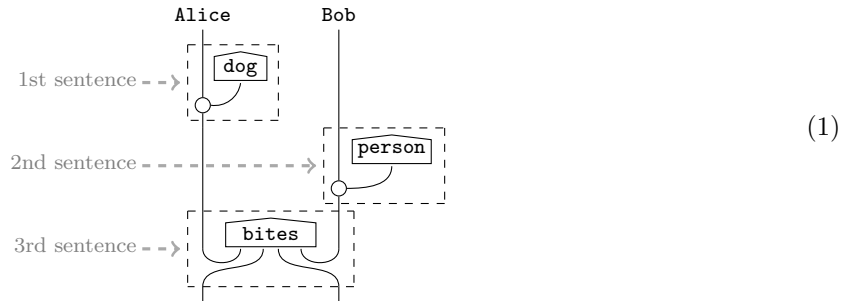
So in particular, the nouns **Alice** and **Bob** are now not states but wires (a.k.a types) and the sentence is an I/O-box with the nouns **Alice** and **Bob** both as inputs and as outputs. In this way, the sentence can act on the nouns and update their meanings. Hence:

A sentence is not a state, but a process,

that represents how words in it are updated by that sentence. The wire-representation of **Alice is a dog** becomes:



Altogether the composition of the three sentences now becomes:



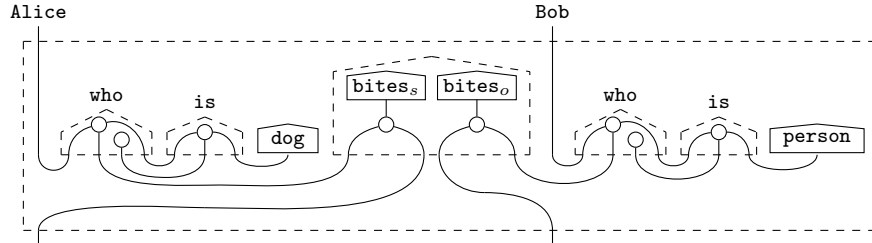
So in general, given text, we end up with a wire diagram that looks like this:



where the sentences themselves also have a wire diagram. In particular, it's a process, and this process alters our understanding of words in the text. This yields another slogan:

Text is a process that alters our understanding of words

The single sentence:



results in the wire-diagram (1), so diagrams can expose that different texts have the same meaning.

A Density Matrix Model

As we are dealing with updating and corresponding information gains, the vector space model of NLP won't do. Density matrices do have a clear notion of information gain, so provide a natural model for DisCoCirc. The main role of the merge-spider is to assign properties. One could set:

$$\begin{array}{c} \boxed{x} \\ | \\ \circ \end{array} := P_x \circ - \circ P_x \tag{3}$$

where P_x is the orthogonal projector on the subspace corresponding to property x . It is indeed natural to think of subspaces as properties, just like in Birkhoff-von Neumann quantum logic. This operation is not commutative, but that also makes perfect sense when thinking of the changing colours of a chameleon, where post-composition should discard previous colours.

Scaled projectors are special cases of density matrices, and this allows us to directly interpret the dot-multiplication in (3) as a multiplication on density matrices as follows:

$$\begin{array}{c} \text{input 2} \\ \text{input 1} \\ \circ \end{array} := - \circ - \circ - \tag{4}$$

When extracting meanings empirically, we may get a proper density matrix ρ_x rather than a projector P_x as a representation of adjectives. Conceptually, this can be interpreted as there being a lack of knowledge about the actual adjective. To see this, just consider the spectral decomposition $\rho_x = \sum_i p_i P_i$ where the P_i are the adjectives and p_i their relative weights. In this case (3) naturally generalises to:

$$\begin{array}{c} \boxed{x} \\ | \\ \circ \end{array} := \sum_i p_i P_i \circ - \circ P_i$$

and there also is a corresponding analogue to (4).