

One learning algorithm

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The one learning algorithm is a concept in machine learning and neuroscience that has been studied extensively by Jeff Hawkins. In his book, *On Intelligence*, he laid out a memory-prediction framework depicting what he thinks the one learning algorithm would consist of. His framework is based upon the work on columnar organization of the cerebral cortex done by Vernon Mountcastle who shows that “all regions of the cortex [perform] the same operation” (Hawkins, 51), and that “there is a common function, a common algorithm, that is performed by all the cortical regions” (Hawkins, 51). In essence, the one learning algorithm accurately predicts outcomes of future input sequences based on patterns that this algorithm learns from regardless of what type of patterns they are (patterns can be images, videos, smell, etc.).

Hawkins is an electrical engineer, and hasn’t had any professional experience in neuroscience. His framework approaches the problem from an engineer’s perspective as well as his personal study of the research done on the cerebral cortex to formulate his framework. He treats the network of neurons in our brain as an encoding problem, and he quotes “All your brain knows is patterns” (Hawkins, 56), which is truly how a mathematician would see it as patterns imply that some formulation could be found to connect the pattern together. He explains that our understanding of the world is based upon patterns and “[correct] predictions result in understanding. Incorrect predictions result in confusion and prompt you to pay attention” (Hawkins, 89).

Hawkins key aim is to create an algorithm that mirrors how we learn, and form what he calls ‘true intelligence’ rather than forming ‘artificial intelligence’. He defines true intelligence as a “[measure of the] capacity to remember and predict patterns in the world, including language, mathematics, physical properties of objects, and social situations” (Hawkins, 97). Hawkins’ view of how our brain recognizes patterns and learns is that our “brain doesn’t “compute” the answers to problems; it retrieves the answers from memory ... entire cortex is a memory system. It isn’t a computer at all” (Hawkins, 68). Thus, Hawkins sees the brain as being able to predict the answer or the best outcome of a certain task by having a very efficient prediction system that utilizes the connection of neurons in our brain to find the best prediction. This is in contrast to how most individuals think about learning and how the neurons in the brain responses to actions. Most individuals believe that the reactions to actions are, in a sense, computed at that particular moment and our brain is built on logic, assumptions, and prior knowledge. This contrast arises the question is learning something we can model as patterns? Can how we think be quantified and put into a series of steps or is it more complex? Can the one learning algorithm be used to represent ‘true intelligence’, and the process of learning in a similar way that our brains do?

Approaching this algorithm from an mathematical perspective, a mathematical model of learning would ignore a lot of things that we believe contribute to learning, and our definition of learning. Lets take an objective definition of learning to be the “process of acquiring modifications in existing knowledge, skills, habits, or tendencies through experience, practice, or exercise” (Merriam-Webster, “Learning”). Knowledge to us is an understanding of pieces of information through experience or learning, and we usually recall these pieces of information using languages that we know, or visual descriptions that we collect. However, to a computer knowledge is represented as a series of numerical val-

ues and things a computer learns are converted, through some measure, into numerical values to be processed through the algorithm and then stored. Without representing them as numerical values there is no way a computer can understand their meanings, as these algorithms require some medium to compare and relate two pieces of information and to classify this new piece of information. Once we input a phrase into this particular learning algorithm it becomes arbitrary. It gets converted into a series of values that the computer performs actions upon to match the particular pattern, and to try to classify it. In contrast, our brains understand this information as it has complex biological traits that allow it to interpret and break down these words, and processes them through its existing knowledge. The question arises that can a learning algorithm react to situations using true intelligence?

If we define true intelligence to be “the ability to learn or understand things or to deal with new or difficult situations” (Merriam-Webster, “Intelligence”) then we can begin to compare it to Hawkins definition of true intelligence, and evaluate if his definition can be used to represent true intelligence. Just concentrating on Hawkins’ definition of intelligence we can observe that it is a more mathematical approach to think about intelligence. Hawkins concentrates more on things that can be quantified such as capacity, and prediction of patterns whereas the other definition concentrates on the ability and the understanding. A learning algorithm doesn’t have the ability to understand the information that’s given, but only the ability to use it to predict using an existing piece of information that is most similar to it by some index.

In addition, in the definition of learning there was an aspect of experience. What we perceive as experience and what the one learning algorithm would define to be experience would be very different. In regards to learning, experience is knowing the prerequisites to the topic at hand or having had experience and some existing knowledge on the topic. To

Hawkins experience is

In conclusion, given Hawkins definition of intelligence the learning capabilities of a computer could never will truly intelligent, but be artificially intelligent. They are being developed by quantifying learning and intelligence and using those arbitrary values to make predictions about the future.

References

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