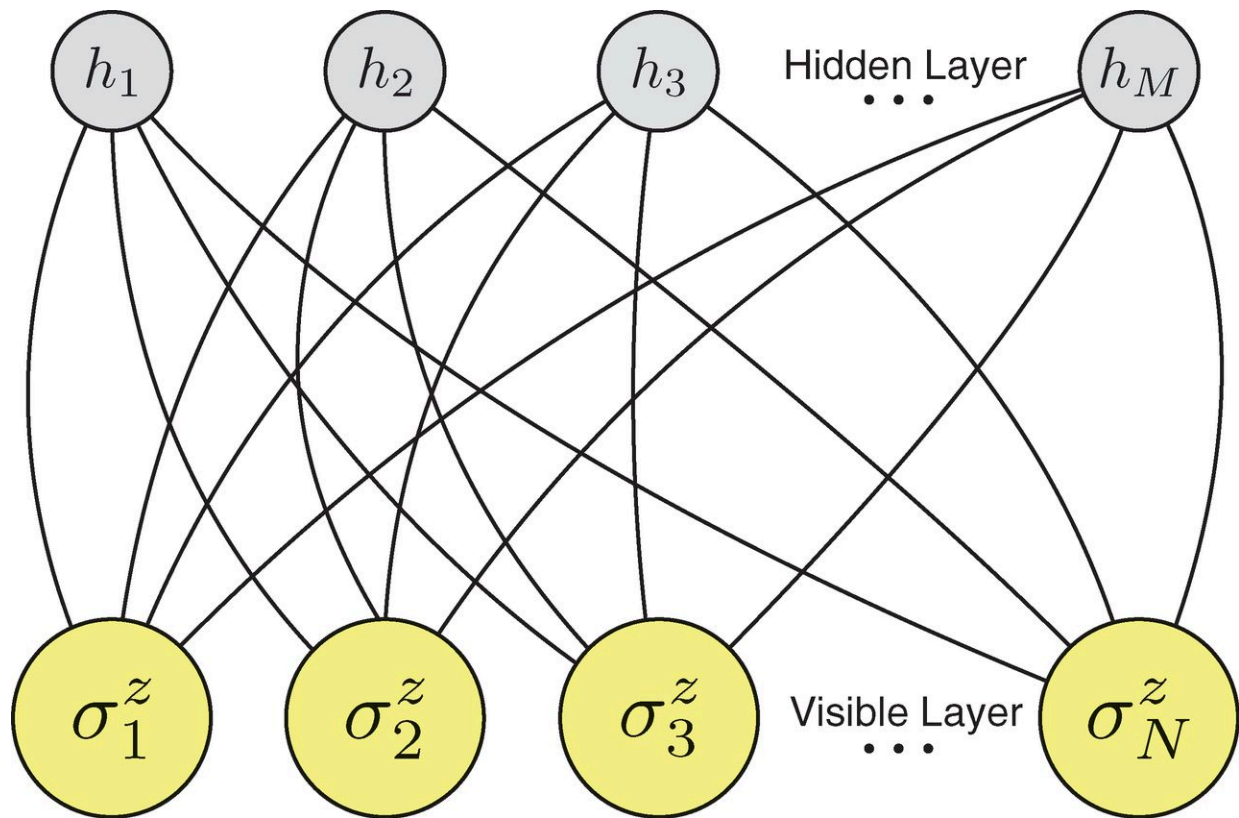


Researchers use artificial neural network to simulate a quantum many-body system

February 10 2017, by Bob Yirka



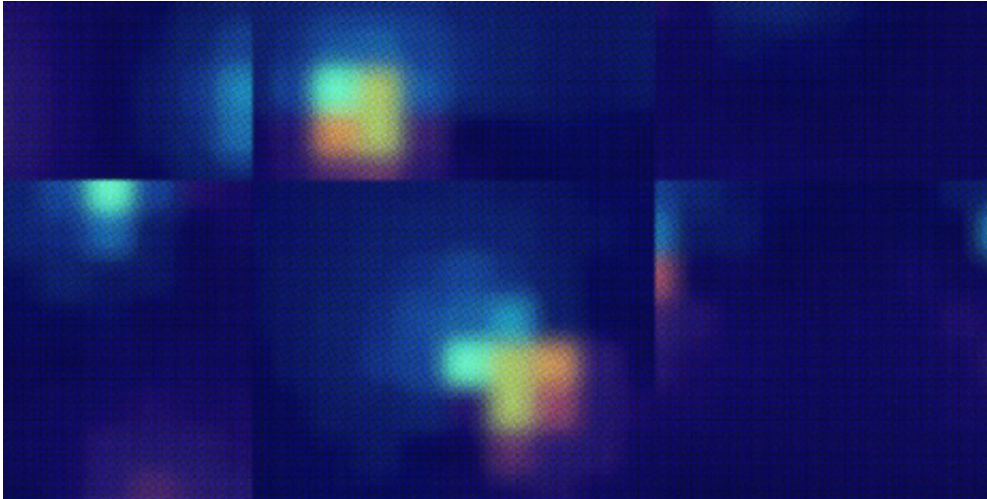
Artificial neural network encoding a many-body quantum state of N spins.
 Credit: (c) *Science* 10 Feb 2017: vol. 355, Issue 6325, pp. 602-606

(Phys.org)—A pair of physicists with ETH Zurich has developed a way to use an artificial neural network to characterize the wave function of a quantum many-body system. In their paper published in the journal

Science, Giuseppe Carleo and Matthias Troyer describe how they coaxed a neural network to simulate some aspects of a quantum many-body system. Michael Hush with the University of New South Wales offers a Perspectives [piece](#) on the work done by the pair in the same journal issue and also outlines the problems other researchers have faced when attempting to solve the same problem.

One of the difficult challenges facing physicists today is coming up with a way to simulate quantum many-body systems, i.e., showing all the states that exist in a given system, such as a chunk of matter. Such systems grow complicated quickly—a group of just 100 quantum particles, for example, could have as many as 10^{35} spin states. Even the most powerful modern computers very quickly become overwhelmed trying to depict such systems. In this new effort, the researchers took a different approach—instead of attempting to calculate every possible state, they used a neural network to generalize the entire system.

The pair began by noting that the system used to defeat a Go world champion last year might be modified in a way that could simulate a many-body system. They created a simplified version of the same type of neural network and programmed it to simulate the [wave function](#) of a multi-body system (by using a set of weights and just one layer of hidden biases). They then followed up by getting the [neural network](#) to figure out the ground state of a system. To see how well their system worked, they ran comparisons with problems that have already been solved and report that their system was better than those that rely on a brute-force approach.



The neural network detects specific patterns in the quantum system. In this case, the network correctly recognises that atoms with an opposite spin tend to pair up. Credit: ETH Zurich / G. Carleo

The system was a proof-of-concept rather than an actual tool for use by physicists, but it demonstrates what is possible—large efforts, as Hush notes, that involve more hidden biases and weights could result in a tool with groundbreaking applications.

More information: Solving the quantum many-body problem with artificial neural networks, *Science* 10 Feb 2017: vol. 355, Issue 6325, pp. 602-606 [science.sciencemag.org/cgi/doi ... 1126/science.aag2302](https://science.sciencemag.org/cgi/doi/10.1126/science.aag2302)

Abstract

The challenge posed by the many-body problem in quantum physics originates from the difficulty of describing the nontrivial correlations encoded in the exponential complexity of the many-body wave function. Here we demonstrate that systematic machine learning of the wave function can reduce this complexity to a tractable computational form for some notable cases of physical interest. We introduce a variational representation of quantum states based on artificial neural networks with

a variable number of hidden neurons. A reinforcement-learning scheme we demonstrate is capable of both finding the ground state and describing the unitary time evolution of complex interacting quantum systems. Our approach achieves high accuracy in describing prototypical interacting spins models in one and two dimensions.

[Press release](#)

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