The Use of Restricted Boltzmann Machines for Modeling a Many-body Quantum System



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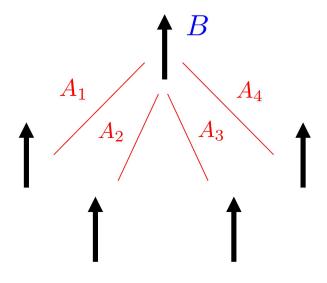
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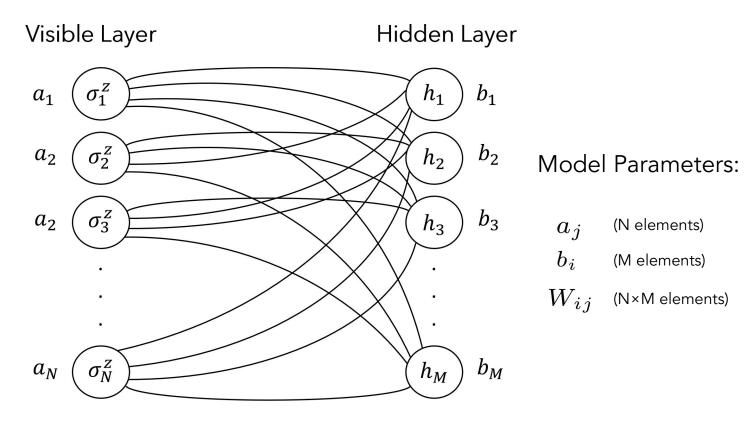
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Restricted Boltzmann Machine

$$H = \mathbf{B}S_0^z + \sum_{k=1}^{N-1} \mathbf{A}_k \mathbf{S}_0 \cdot \mathbf{S}_k$$



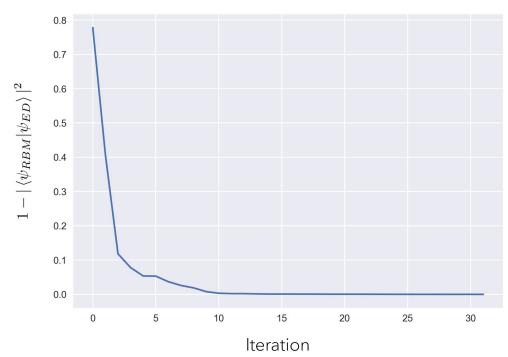


$$\Psi(S; \mathbf{a}, \mathbf{b}, \mathbf{W}) = \sum_{\{h_i\}} e^{\sum_j a_j \sigma_j^z + \sum_i b_i h_i + \sum_{ij} W_{ij} h_i \sigma_j^z}$$

Ground State Determination

$$E(\mathbf{a}, \mathbf{b}, \mathbf{W}) = \frac{\langle \psi_{RBM} | H | \psi_{RBM} \rangle}{\langle \psi_{RBM} | \psi_{RBM} \rangle}$$





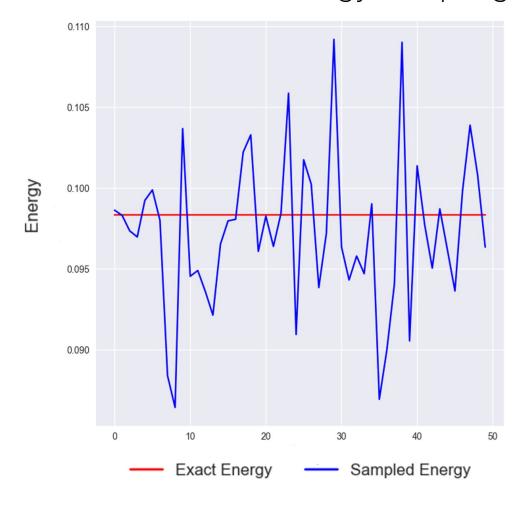
- \bullet The variational energy is minimal when ψ_{RBM} accurately models the ground state
- Learning is achieved through this minimization
- RBM is sufficiently expressive to model the ground state

Network Training

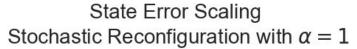
- Stochastic Reconfiguration [1]
- Monte Carlo Sampling [1]

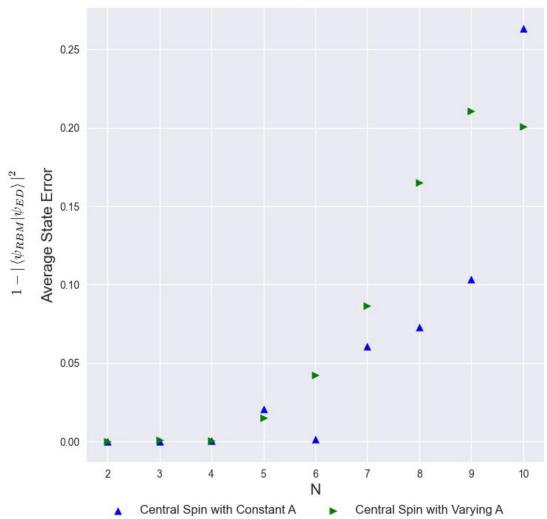
$$\begin{split} \langle \hat{H} \rangle &= \frac{\sum_{\sigma,\sigma'} \Psi^*(\sigma) \langle \sigma | \hat{H} | \sigma' \rangle \Psi(\sigma')}{\sum_{\sigma} |\Psi(\sigma)|^2} \\ &= \sum_{\sigma} \left(\sum_{\sigma'} \langle \sigma | \hat{H} | \sigma' \rangle \frac{\Psi(\sigma')}{\Psi(\sigma)} \right) \frac{|\Psi(\sigma)|^2}{\sum_{\sigma'} |\Psi(\sigma')|^2} \\ &\approx \left\langle \sum_{\sigma'} \langle \sigma | \hat{H} | \sigma' \rangle \frac{\Psi(\sigma')}{\Psi(\sigma)} \right\rangle_{\sigma} \end{split}$$

Monte Carlo Energy Sampling



Hamiltonian Comparison





$$H = \mathbf{B}S_0^z + \sum_{k=1}^{N-1} \mathbf{A}_k \mathbf{S}_0 \cdot \mathbf{S}_k$$

Constant coupling

$$A_k = 1 \ \forall k, B = 1$$

Varying coupling

$$A_k = \frac{A}{N_0} e^{\frac{-k}{N_0}}$$

$$A = \frac{N}{2}, B = \frac{N}{2}, N_0 = \frac{N}{2}$$

References

1. G. Carleo and M. Troyer, Solving the quantum many-body problem with artificial neural networks, Science 355, 602 (2017)