

Optimization by Mean Field Annealing

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Abstract: Nearly optimal solutions to many combinatorial problems can be found using stochastic simulated annealing. This paper extends the concept of simulated annealing from its original formulation as a Markov process to a new formulation based on mean field theory. Mean field annealing essentially replaces the discrete degrees of freedom in simulated annealing with their average values as computed by the mean field approximation. The net result is that equilibrium at a given temperature is achieved 1-2 orders of magnitude faster than with simulated annealing. A general framework for the mean field annealing algorithm is derived, and its relationship to Hopfield networks is shown. The behavior of MFA is examined both analytically and experimentally for a generic combinatorial optimization problem: graph bipartitioning. This analysis indicates the presence of critical temperatures which could be important in improving the performance of neural networks.

STOCHASTIC VERSUS MEAN FIELD

In combinatorial optimization problems, an objective function or Hamiltonian, $H(s)$, is presented which depends on a vector of interacting spins, $S = \{s_1, \dots, s_N\}$, in some complex nonlinear way. Stochastic simulated annealing (SSA) (S. Kirkpatrick, C. Gelatt, and M. Vecchi (1983)) finds a global minimum of H by combining gradient descent with a random process. This combination allows, under certain conditions, choices of s which actually increase H , thus providing SSA with a mechanism for escaping from local minima. The frequency and severity of these uphill moves is reduced by slowly decreasing a parameter T (often referred to as the temperature) such that the system settles into a global optimum. Two conceptual operations are involved in simulated annealing: a thermodynamic operation which schedules decreases in the temperature, and a relaxation operation.