

TUTORIAL 08

7CCMCS04

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✎ Problem 8.1 A neural network of N neurons $\sigma_i = \pm 1$ $i = 1, \dots, N$ evolves according to parallel Glauber dynamics, so that the probability $P_{\sigma}(t)$ to find the system in configuration $\sigma = (\sigma_1, \dots, \sigma_N)$ evolves according to the Markov chain

$$P_{\sigma}(t+1) = \sum_{\sigma'} Q_{\sigma\sigma'} P_{\sigma'}(t)$$

with

$$Q_{\sigma\sigma'} = \prod_{i=1}^N \frac{e^{\beta\sigma_i h_i(\sigma')}}{2 \cosh(\beta h_i(\sigma'))}, \quad h_i(\sigma) = \sum_j J_{ij} \sigma_j + \theta_i.$$

- a. Show that $\sum_{\sigma} Q_{\sigma\sigma'} = 1$.
- b. Show that for symmetric interactions $J_{ij} = J_{ji}$ the system satisfies detailed balance with the Boltzmann distribution

$$\Pi_{\sigma} = \frac{1}{Z} e^{-\beta H(\sigma)} \quad Z = \sum_{\sigma} e^{-\beta H(\sigma)},$$

with

$$H(\sigma) = - \sum_i \theta_i \sigma_i - \beta^{-1} \sum_i \ln \cosh(\beta h_i(\sigma)).$$

✎ Problem 8.2 A neural network of N neurons $\sigma_i = \pm 1$, $i = 1, \dots, N$ evolves according to sequential Glauber dynamics so that at every time step $\tau = N^{-1}$ a spin σ_i is picked at random and updated according to the stochastic rule

$$\sigma_i(t+\tau) = \begin{cases} -\sigma_i(t) & \text{with probability } W_i(\sigma(t)) \\ \sigma_i(t) & \text{with probability } 1 - W_i(\sigma(t)) \end{cases}$$

where

$$W_i(\sigma) = \frac{1 - \sigma_i \tanh \beta h_i(\sigma)}{2}$$

and we have denoted $\beta = T^{-1}$ the inverse temperature and h_i the effective local field

$$h_i(\sigma) = \sum_{j \neq i} J_{ij} \sigma_j.$$

- (a) Show that for $N \gg 1$ the microstate distribution $P_{\sigma}(t)$ evolves according to the master equation

$$\dot{P}_{\sigma}(t) = \sum_i (W_i(F_i \sigma) P_{F_i \sigma}(t) - W_i(\sigma) P_{\sigma}(t)),$$

where F_i is the i -spin flip operator $F_i \sigma = (\sigma_1, \dots, \sigma_{i-1}, -\sigma_i, \sigma_{i+1}, \dots, \sigma_N)$.

- (b) Prove that the Glauber rates $W_i(\sigma)$ satisfy detailed balance with the Gibbs–Boltzmann distribution

$$\Pi_{\sigma} = \frac{e^{-\beta H(\sigma)}}{\sum_{\sigma'} e^{-\beta H(\sigma')}}.$$

with

$$H(\sigma) = -\frac{1}{2} \sum_{k \neq \ell} J_{kl} \sigma_k \sigma_{\ell}$$

for symmetric synaptic interaction without self-interactions, i.e., $J_{ii} = 0 \forall i$ and $J_{ij} = J_{ji} \forall i, j$.
