MPF objective function for an Ising model

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This document derives the MPF objective function for the case of an Ising model, and connections to all states which differ by a single bit flip. It is written quickly, and not well proofread!! Typos and missing steps are likely...

For the Ising model, the energy function is

$$E = \mathbf{x}^T \mathbf{J} \mathbf{x} \tag{1}$$

where $\mathbf{x} \in \{0,1\}^N$, $\mathbf{J} \in \mathcal{R}^{N \times N}$, and \mathbf{J} is symmetric ($\mathbf{J} = \mathbf{J}^T$). The MPF objective function is

$$K = \sum_{\mathbf{x} \in \mathcal{D}} \sum_{n} \exp \left(\frac{1}{2} \left[E(\mathbf{x}) - E(\mathbf{x} + \mathbf{d}(\mathbf{x}, n)) \right] \right)$$
 (2)

where the sum over n indicates a sum over all data dimensions, and the function $\mathbf{d}(\mathbf{x},n) \in \{-1,0,1\}^N$ is

$$\mathbf{d}(\mathbf{x}, n)_i = \begin{cases} 0 & i \neq n \\ -(2x_i - 1) & i = n \end{cases}$$
 (3)

For the Ising model, the MPF objective function becomes

$$K = \sum_{\mathbf{x} \in \mathcal{D}} \sum_{n} \exp \left(\frac{1}{2} \left[\mathbf{x}^{T} \mathbf{J} \mathbf{x} - (\mathbf{x} + \mathbf{d}(\mathbf{x}, n))^{T} \mathbf{J} (\mathbf{x} + \mathbf{d}(\mathbf{x}, n)) \right] \right)$$
(4)

$$= \sum_{\mathbf{x} \in \mathcal{D}} \sum_{n} \exp \left(\frac{1}{2} \left[\mathbf{x}^{T} \mathbf{J} \mathbf{x} - \left(\mathbf{x}^{T} \mathbf{J} \mathbf{x} + 2 \mathbf{x}^{T} \mathbf{J} \mathbf{d} (\mathbf{x}, n) + \mathbf{d} (\mathbf{x}, n)^{T} \mathbf{J} \mathbf{d} (\mathbf{x}, n) \right) \right] \right)$$
(5)

$$= \sum_{\mathbf{x} \in \mathcal{D}} \sum_{n} \exp \left(-\frac{1}{2} \left[2\mathbf{x}^{T} \mathbf{J} \mathbf{d}(\mathbf{x}, n) + \mathbf{d}(\mathbf{x}, n)^{T} \mathbf{J} \mathbf{d}(\mathbf{x}, n) \right] \right)$$
(6)

$$= \sum_{\mathbf{x} \in \mathcal{D}} \sum_{n} \exp\left(-\frac{1}{2} \left[2 \sum_{i} x_{i} J_{in} \left(1 - 2x_{n}\right) + J_{nn} \right] \right)$$
 (7)

$$= \sum_{\mathbf{x} \in \mathcal{D}} \sum_{n} \exp\left(\left[(2x_n - 1)\sum_{i} x_i J_{in} - \frac{1}{2} J_{nn}\right]\right). \tag{8}$$

Assume the symmetry constraint on \mathbf{J} is enforced by writing it in terms of a another possibly asymmetric matrix $\mathbf{J}' \in \mathcal{R}^{N \times N}$,

$$\mathbf{J} = \frac{1}{2}\mathbf{J}' + \frac{1}{2}\mathbf{J}'^T. \tag{9}$$

The derivative of the MPF objective function with respect to \mathbf{J}' is

$$\frac{\partial K}{\partial J'_{lm}} = \frac{1}{2} \sum_{\mathbf{x} \in \mathcal{D}} \exp\left(\left[\left(2x_m - 1\right) \sum_{i} x_i J_{im} - \frac{1}{2} J_{mm}\right]\right) \left[\left(2x_m - 1\right) x_l - \delta_{lm} \frac{1}{2}\right] + \frac{1}{2} \sum_{\mathbf{x} \in \mathcal{D}} \exp\left(\left[\left(2x_l - 1\right) \sum_{i} x_i J_{il} - \frac{1}{2} J_{ll}\right]\right) \left[\left(2x_l - 1\right) x_m - \delta_{ml} \frac{1}{2}\right],$$
(10)

where the second term is simply the first term with indices l and m reversed.

Note that both the objective function and gradient can be calculated using matrix operations (no for loops). See the code.