

## Variational Inference (11/4/13)

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### 1 Introduction

### 2 Ising Model

Before proceeding with variational inference, it is helpful to review the Ising model. The main idea behind the Ising model is a lattice of unobserved variables  $(x_1, \dots, x_n)$ , each with its own (noisy) observation  $(y_1, \dots, y_n)$ . For example, we can think of the lattice as pixels in a black and white image ( $x_i \in \{-1, 1\}$ ), with a noisy grayscale observation of the pixels ( $y_i \in R$ ). Our goal in this case would be to obtain a de-noised version of the image. More generally, we wish to draw inferences about the unobserved lattice  $X$  from the observed values  $Y$ . Figure 1 illustrates an Ising model for  $n = 9$ , with the latent nodes colored white and the observed nodes shaded grey.

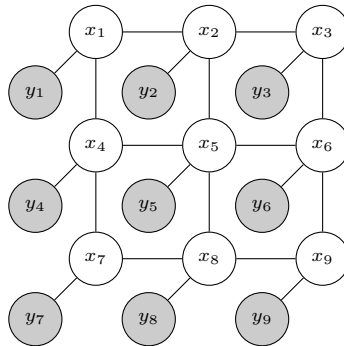


Figure 1: Schematic of an Ising Model

We now define the potential functions of the Ising model:

$$\begin{aligned}\psi_s(x_s) &= p(y_s|x_s) \equiv L_s(x_s) \\ \psi_{st}(x_s x_t) &= W_{st} x_s x_t\end{aligned}$$

Continuing with our image example above, we could set  $W_{st} = 1$ . In general, we set  $W$  to positive values if we want neighbors to agree, and negative values if we want them to differ.

Now we can specify our priors. Let  $N(i)$  be a function that returns the first-degree neighbors of node  $i$ . For example, in Figure 1, calling  $N(x_1)$  would return nodes  $x_2$  and  $x_4$ .

$$p(x) = \frac{1}{z_0} \exp\left\{-\sum_{i=1}^n \sum_{j \in N(i)} x_i x_j\right\}$$

$$\begin{aligned}
p(y|x) &= \prod_{i=1}^n \exp\{-L_i(x_i)\} \\
p(x|y) &= \frac{1}{z} \exp\left\{-\sum_{i=1}^n \sum_{j \in N(i)} x_i x_j - \sum_{i=1}^n L_i(x_i)\right\}
\end{aligned}$$

## 2.1 Mean Field Version of the Ising Model

Having seen an example of a basic Ising model, we now turn our attention to how we can analyze the mean field version of such a model. We do this by “breaking” the edges between the latent variables, and assigning a mean parameter  $\mu$  to each of them. The new structure is illustrated in Figure 2, with the same color coding as above.

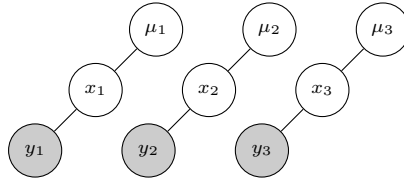


Figure 2: Mean Field Version of an Ising Model

## 3 Loopy Belief Propagation