Probabilistic Graphical Models Exercise Sheet No. 3

Due Date: November 30th, 10 am

Hand in: By e-mail to (jlange[at]mpi-inf.mpg.de). Your submission must include your source code files. Begin the subject of your e-mail with [pgm].

1 Binary Image Denoising

Points: 3

In binary image denoising we are given an observed black-and-white noisy image y and want to recover the original uncorrupted image x as good as possible. This can modeled as an undirected graphical model with random variables $x_i, y_i \in \{-1, 1\}$ for each pixel i = 1, ..., N, where $x_i = 1$ means pixel i is white. The x_i are arranged in a grid graph and each y_i is connected by an edge with x_i . The joint distribution $p(x, y) \propto \exp(-E(x, y))$ is described by the energy function

$$E(x,y) = h \sum_{i=1}^{N} x_i - \beta \sum_{i \sim j} x_i x_j - \eta \sum_{i=1}^{N} x_i y_i.$$
 (1)

The constants $\beta, \eta \ge 0, h \in \mathbb{R}$ are model parameters and $i \sim j$ denotes that the variables x_i and x_j are horizontally/vertically adjacent.

a) What are the roles of the three terms hx_i , $-\beta x_i x_j$ and $-\eta x_i y_i$? For each of them explain how a high (respectively low) parameter value influences the model.

2 Iterated Conditional Modes

Points: 7

Download the noisy image y and the ground truth image x^{GT} from the course website. In order to recover an original image x given a noisy observation y, we need to compute the MAP estimate

$$x^* = \operatorname*{argmax}_{x} p(x \mid y). \tag{2}$$

b) Rewrite (2) as an energy minimization problem.

A simple heuristic method to tackle (2) is the *Iterated Conditional Modes* (ICM) algorithm. It works by iteratively optimizing one variable x_i at a time while keeping all other variables fixed.

- b) Implement Iterated Conditional Modes (ICM) to compute a heuristic solution to (2).
- c) Set the parameters of the model (1) to h = 0.04, $\beta = 0.93$ and $\eta = 0.75$. Compute an estimate for x^* by ICM starting from the noisy image y with an update order of your choice. Your program must report the fraction of incorrectly recovered pixels

$$L(x, x^{\text{GT}}) = \frac{1}{N} \sum_{i=1}^{N} [x_i \neq x_i^{\text{GT}}]$$

for your estimated image x.

2.1 Block-ICM (Bonus Task)

Points: 4

- d) Implement Block-ICM where each block is a single row (or column) of the image. For this you need to implement an exact solver for the energy minimization problem on chain graphs (dynamic programming).
- e) Repeat task c) with your Block-ICM algorithm.

3 Max Product Linear Programming

Points: 6

- d) Implement Max Product Linear Programming (MPLP) to solve the dual problem corresponding to (2).
- e) Apply your algorithm to the noisy image y with the same parameter settings as in c). Reconstruct images from your dual solution by naïve rounding and by subsequently running ICM (initialized with your rounding solution). Report the primal-dual gap of your solutions. Compare the loss $L(x, x^{\text{GT}})$ to your results from the previous task.