

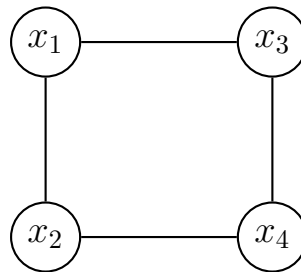
# CS 446: Machine Learning

## Homework 7

Due on Tuesday, March 6, 2018, 11:59 a.m. Central Time

### 1. [4 points] Inference Methods for Discrete Markov Random Fields

For this problem, consider the following Markov Random Field, where each node can be assigned one of the values in  $\{1, 2, 3, 4, 5\}$ :



- (a) To conduct MAP inference on this graph using exhaustive search, how many configurations must be checked?

**Solution:**  $5^4 = 625$  (1 pt)

- (b) Can MAP inference be run on this graph using a dynamic programming algorithm? Why or why not?

**Solution:** No. The graph is not a tree. (2 pts, one for "No" and one for explanation)

- (c) To run MAP inference on this graph using loopy belief propagation, how many messages must be computed?

**Solution:** # Messages = Number of nodes  $\times$  neighbors per node  $\times$  number of node assignments  
 $= 4 \times 2 \times 5$   
 $= 40$  (1 pt)

### 2. [7 points] ILP Inference formulation in Discrete Markov Random Fields

- (a) Suppose we have two variables  $x_1 \in \{0, 1\}$  and  $x_2 \in \{0, 1\}$  and their local evidence functions  $\theta_1(x_1)$  and  $\theta_2(x_2)$  as well as a pairwise function  $\theta_{1,2}(x_1, x_2)$ . Using this setup, inference solves  $\arg \max_{x_1, x_2} \theta_1(x_1) + \theta_2(x_2) + \theta_{1,2}(x_1, x_2)$ . Using

$$\theta_1(x_1) = \begin{cases} 1 & \text{if } x_1 = 0 \\ 2 & \text{otherwise} \end{cases} \quad \theta_2(x_2) = \begin{cases} 1 & \text{if } x_2 = 0 \\ 2 & \text{otherwise} \end{cases}$$

$$\theta_{1,2}(x_1, x_2) = \begin{cases} -1 & \text{otherwise} \\ 2 & \text{if } x_1 = 0 \text{ \& } x_2 = 1 \end{cases}$$

what is the integer linear programming formulation of the inference task? Make the cost function and constraints explicit for the given problem, i.e., do not use a general formulation.

**Solution:**

$$\begin{aligned} \max_b \quad & \begin{bmatrix} 1 \\ 2 \\ 1 \\ 2 \\ -1 \\ -1 \\ 2 \\ -1 \end{bmatrix}^\top \begin{bmatrix} b_1(0) \\ b_1(1) \\ b_2(0) \\ b_2(1) \\ b_{1,2}(0,0) \\ b_{1,2}(1,0) \\ b_{1,2}(0,1) \\ b_{1,2}(1,1) \end{bmatrix} \\ \text{s.t.} \quad & \begin{cases} b_1(0), b_1(1), b_2(0), b_2(1), b_{1,2}(0,0), b_{1,2}(1,0), b_{1,2}(0,1), b_{1,2}(1,1) \in \{0, 1\} \\ b_1(0) + b_1(1) = 1, b_2(0) + b_2(1) = 1 \\ b_{1,2}(0,0) + b_{1,2}(1,0) + b_{1,2}(0,1) + b_{1,2}(1,1) = 1 \\ b_1(0) = b_{1,2}(0,0) + b_{1,2}(0,1) \\ b_1(1) = b_{1,2}(1,0) + b_{1,2}(1,1) \\ b_2(0) = b_{1,2}(0,0) + b_{1,2}(1,0) \\ b_2(1) = b_{1,2}(0,1) + b_{1,2}(1,1) \end{cases} \end{aligned}$$

(4 pts - one for max, one for  $b_r \in \{0, 1\}$ , one for intra-region marginalization constraints, one for inter-region marginalization constraints)

(b) What is the solution (value and argument) to the program in part (a).

**Solution:**

argument:  $b_1(0) = 1; b_1(1) = 0; b_2(0) = 0; b_2(1) = 1; b_{1,2}(0,0) = 0; b_{1,2}(1,0) = 0; b_{1,2}(0,1) = 1; b_{1,2}(1,1) = 0$ ; value: 5 (2 pts - one for value, one for argument)

(c) Why do we typically not use the integer linear program for reasonably sized MRFs?

**Solution:**

It can be very slow (ILP is NP-Hard). (1 pt)