

## CS 228T QUIZ 1

1. The idea behind basic Monte Carlo integration is to replace intractable expectations  $\mathbb{E}_p[f(X)]$  with sums  $\sum_{i=1}^n f(x_i)$ , where  $x_i$  are independent samples from  $p(x)$ . What potential difficulty in this basic method does Markov chain Monte Carlo address?

2. Consider the following two statements:

- i. A Markov chain with transition operator  $T$  satisfies detailed balance with respect to a distribution  $p$ , *i.e.*,

$$p(x)T(x \rightarrow x') = p(x')T(x' \rightarrow x)$$

holds for all possible outcomes  $x, x'$ .

- ii. The distribution  $p$  is a stationary distribution for  $T$ .

Assume here that the Markov chains mentioned are regular and have finite state space, so they can get from any state to any other state in a finite number of steps with positive probability.

Which of the following is true?

- (a) (i) implies (ii), but not vice versa
- (b) (ii) implies (i), but not vice versa
- (c) (i) if and only if (ii)
- (d) (i) and (ii) are unrelated

3. Consider the Bayesian network  $X \rightarrow Y \rightarrow Z$ . If the current sample is  $(x_0, y_0, z_0)$ , and the first substep of Gibbs sampling is to sample  $y$ , with what probability will the first subsample be  $(x_0, y_1, z_0)$ ?

- (a)  $p(y_1 \mid x_0)$
- (b)  $p(y_1 \mid x_0, z_0)$
- (c)  $p(y_1 \mid z_0)$
- (d)  $p(y_1)$

4. Suppose you have a bipartite MRF with two sets of variables,  $\{X_1, \dots, X_n\}$  and  $\{Y_1, \dots, Y_n\}$ . Assume that  $n$  is large, e.g.  $n = 1000$ . Each  $X_i$  is connected to each  $Y_j$ , none of the  $Y_j$ 's are connected to each other, and the  $X_i$ 's are internally connected using a tree structure. Assume that the edges in the tree structure connecting the  $X_i$ 's induce very strong correlations between the  $X_i$ 's that they connect. If you are applying collapsed Gibbs sampling to this MRF, which variables should you use as the sampled variables?

- (a) The  $X_i$ 's.
- (b) The  $Y_j$ 's.
- (c) All the variables.
- (d) Either the  $X_i$ 's or the  $Y_j$ 's, it doesn't matter.

5. A difficulty with MCMC methods is that it can be difficult to know when the chain has mixed. There are a number of heuristic strategies that attempt to mitigate this problem. Which of the following is generally a good default approach? (Again, assume we are sampling in high dimensions and that the state space is very large.)

- (a) Run one chain for a long time.
- (b) Run a large number of chains for a short time, each initialized at different starting values.
- (c) Run a few chains for a medium time, each initialized at different starting values.

6. When running an MCMC method, is it better to have to sample more variables or fewer variables?