CS60073: Advanced Machine Learning

End-Autumn Semester Exam

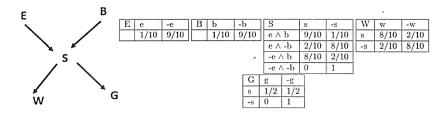
Time: 3:00hrs Answer all FOUR questions. Max mark: $25 \times 4 = 100$

1.A. Consider the Markov chain with three states, $S = \{1, 2, 3\}$, that has the following transition matrix.

$$T = \begin{bmatrix} 1/2 & 1/4 & 1/4 \\ 1/3 & 0 & 2/3 \\ 1/2 & 1/2 & 0 \end{bmatrix}$$

Draw the state transition diagram for this chain. Let X_t denote the state at time t. If we know $P(X_1 = 1) = P(X_1 = 2) = 1/4$, find $P(X_1 = 3, X_2 = 2, X_3 = 1)$. [10]

- 1.B. We would like to sample from a distribution $p(x,y) = \exp(-x^2 (y-x^2)^2)$, where $x,y \in \mathbb{R}$. We have access to a Gaussian and an uniform random number generator. [15]
- (i) Suggest a proposal distribution and derive the acceptance probability for the Metropolis-Hastings sampler that might be used for this purpose.
- (ii) Write down the steps of Gibbs sampling of p(x, y). Derive the conditional probability p(y|x) to be used in the Gibbs sampler.
- (iii) Describe details of a rejection sampling technique for p(y|x) in the Gibbs sampler.
- 2A. Define the Kullback-Leibler divergence between two distributions p and q. [5]
- 2.B. Pose variational inference on distributions p and q, involving observed variable X and fatent variable Z, as an optimization problem. [5]
- 2.C. Show that the evidence lower bound ELBO $\mathcal{L}(q(Z)) = \int q(Z) \log \frac{p(X,Z)}{q(Z)} dZ$ is indeed a lower bound of the evidence $\log p(X)$. What is the modified optimization problem
 - obtained due to this result? [10]
 2.D. Clearly state the assumptions made in mean field variational inference. [5]
 - 3. Consider the Bayesian network of five binary variables E, B, S, W, G given below.



- A. Compute the following probabilities: (show your calculation)
- (i) Given that G is true, what is the probability that S is true?
- (H) If B is true and E is false, what is the probability that W is true?
- 3. Does the message passing belief propagation algorithm perform exact inference for this Bayesian network? Explain your answer. [5]
- C. We would like to perform approximate inference by Gibbs sampling for this network. A topologically sorted variable ordering is used. The variable E is observed to be true. What is the probability that -
- The first sample is: E, B, W, G are true, and S is false.
- \mathcal{L}_{ii}) The second sample is: E, B, G are true, and S, W are false.

[P.T.O]

[10]

4.A. What is an independence map (I-Map) for a given distribution p?	[5]
4.B. Draw a Bayesian network which is an I-Map for the distribution $P(A, B)$	$A,\ldots,J) =$
$\mathcal{N}(A C,D,H)P(B D,E,G)P(C F,I)P(D Q,H)P(E J)P(F I)P(G H,J)P(C F,I)P(D Q,H)P(E J)P(E J)P(E I)P(E J)P(E J)P$	$(H)P(\cancel{V})P(J)$
	and H , and
(iii) \mathcal{L} and G .	[10]
4.C. Draw a factor graph representation of the above Bayesian network.	[5]
4D. If possible, draw an undirected Markov Random Field to represent the	e above dis-
tribution. Define appropriate clique potentials.	[5]

— BEST WISHES —