Practice Problems for CSC 412/2506 Midterm

1. Let p(k) be a one-dimensional discrete distribution that we wish to approximate, with support on nonnegative integers. One way to fit an approximating distribution q(k) is to minimize the Kullback-Leibler divergence:

$$KL(p||q) = \sum_{k=0}^{\infty} p(k) \log \frac{p(k)}{q(k)}$$

Show that when q(k) is a Poisson distribution, this KL divergence is minimized by setting λ to the mean of p(k).

$$q(k) = \frac{\lambda^k \exp(-\lambda)}{k!}$$

Solution:

2. Recall that the definition of an exponential family model is:

$$f(x|\eta) = h(x)g(\eta)\exp(\eta^{\top}T(x))$$

where:

- η are the parameters
- T(x) are the sufficient statistics
- h(x) is the base measure
- $g(\eta)$ is the normalizing constant

Consider the univariate Gaussian, with mean μ and precision $\lambda = 1/\sigma^2$ is:

$$p(D|\mu, \lambda = \prod_{i=1}^{N} \left(\frac{\lambda}{2\pi}\right)^{1/2} \exp\left(-\frac{\lambda}{2}(x_i - \mu)^2\right)$$

What are η and T(x) for this distribution when it is represented in exponential family form? Solution: 3. Consider the DAG in Figure 1. List all variables that are independent of A given evidence on B. You can use the "Bayes ball" algorithm, the d-separation criterion, or the method of converting to an undirected graph (all should give the same result). Solution:



Figure 1: A directed graphical model.

- 4. Variable Elimination: Murphy 20.1 Solution:
- 5. Consider a tree-structured factor graph over discrete variables, and suppose we wish to evaluate the joint distribution $[(x_a, x_b)$ associated with two variables x_a and x_b that do not belong to a common factor. Define a procedure for using the sum-product algorithm to evaluate this joint distribution in which one of the variables is successively clamped to each of its allowed values. Solution: