EE3731C Tutorial - Statistical Signal 3.5

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- 1. Let's further motivate why we might want to generate samples from a probability distribution. Let \vec{x} be a random vector consisting of N random variables: $\{x_1, \dots, x_N\}$. Suppose each random variable x_n can take on values from 1 to K. Assume N=20 and K=10.
 - (a) Given an observation y, to find the MAP estimate of \vec{x} given y, we want:

$$\vec{x}_{MAP} = \operatorname*{argmax}_{\vec{x}} p(\vec{x}|y) \tag{1}$$

To evaluate Eq. (1), we can iterate through every possible value of \vec{x} and keep track of the value of \vec{x} with the highest posterior probability. Suppose evaluating $p(\vec{x}|y)$ for a particular pair of \vec{x} and y takes 1ms. How long would this exhaustive search approach take to compute the MAP estimate?

(b) Given an observation y, suppose we want to find the MMSE estimate of \vec{x} given y:

$$\vec{x}^{MMSE} = E_{p(\vec{x}|y)}(\vec{x})$$

We can compute the above MMSE estimate for each x_n independently. In other words

$$x_n^{MMSE} = E_{p(x_n|y)}(x_n) = \sum_{x_n=1}^K x_n p(x_n|y)$$

Suppose evaluating $p(\vec{x}|y)$ for a particular pair of \vec{x} and y takes 1ms. How long would it take to compute x_n^{MMSE} ? How long would it take to compute \vec{x}^{MMSE} ?

2. Suppose we want to sample $x \in \{1, 2, 3, 4\}$,

$$p(x = 1) = 0.4$$

 $p(x = 2) = 0.1$
 $p(x = 3) = 0.2$

$$p(x=4) = 0.3$$

with the Metropolis algorithm. Suppose the proposal distribution is

$$q(x'=x-1|x)=q(x'=x|x)=q(x'=x+1|x)=1/3 \qquad \text{for } x\in\{2,3\}$$

$$q(x'=1|x)=2/3; q(x'=2|x)=1/3 \qquad \text{for } x=1$$

$$q(x'=3|x)=1/3; q(x'=4|x)=2/3 \qquad \text{for } x=4$$

- (a) Suppose current x is 2 and x' is 3, what is the probability of accepting x' as the new x?
- (b) Suppose current x is 3 and x' is 2, what is the probability of accepting x' as the new x?
- (c) What is the equivalent Markov chain transition matrix in this Metropolis algorithm?
- (d) Show that this particular Metropolis algorithm will converge to p(x) by showing that the equivalent markov chain satisfy the conditions of the Fundamental Theorem of Markov Chain.