

# Markov Chain Monte Carlo

ISC 5225 Midterm

Oct 13, 2022

## 1. Transformation Method for Normal Distribution

Suppose we want to use the transformation method to sample from the normal distribution  $p(x; \mu, \sigma^2)$ ,

$$p(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2} \frac{(x - \mu)^2}{\sigma^2}\right), \quad (1)$$

with mean  $\mu = 0$ , and variance  $\sigma^2 = 1$ . The cumulative distribution function corresponding to this unit normal distribution is,

$$F(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-z^2/2} dz = \frac{1}{2} \left[ 1 + \operatorname{erf}\left(\frac{x}{\sqrt{2}}\right) \right], \quad (2)$$

where  $\operatorname{erf}(x)$  is the error function, cannot be analytically inverted to find  $x = F^{-1}(u)$  for use in the transformation method.

Interestingly, a reasonably accurate approximation for  $F^{-1}(\cdot)$  is given by:

$$x = F^{-1}(u) \approx \begin{cases} \frac{a_0 + a_1 t}{1 + b_1 t + b_2 t^2} - t & \text{for } u \leq 0.5 \text{ with } t = \sqrt{-2 \log(u)} \\ t - \frac{a_0 + a_1 t}{1 + b_1 t + b_2 t^2} & \text{for } u > 0.5 \text{ with } t = \sqrt{-2 \log(1 - u)} \end{cases} \quad (3)$$

with constants  $a_0 = 2.30753$ ,  $a_1 = 0.27061$ ,  $b_1 = 0.99229$ , and  $b_2 = 0.04481$ .<sup>1</sup>

- Write a program to sample from  $p(x; \mu = 0, \sigma^2 = 1)$  using this approximation. Generate  $n = 10^4$  samples from this distribution.
- Compare the histogram of these samples with the PDF, eqn (1)
- Compute the mean and standard deviation of these samples.

**Points:** (a) 20, (b) 10, (c) 5

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<sup>1</sup>e.g. say,  $u = 0.4 \leq 0.5$ , then set  $t = \sqrt{-2 \log(0.4)}$ , and substitute into the top branch and return  $x = (a_0 + a_1 t)/(1 + b_1 t + b_2 t^2) - t$ . Follow the lower branch if  $u > 0.5$ .

## 2. Importance Sampling

Consider the integral,

$$I = \int_{-\infty}^{\infty} \frac{\theta}{1 + \theta^2} e^{-(\theta-1)^2/2} d\theta. \quad (4)$$

For this question, you are encouraged to use the sampler developed in the previous question.<sup>2</sup>

- (a) Use importance sampling with  $n = 10^4$  samples drawn from the normal distribution,  $\pi(\theta) = p(\theta; \mu = 0, \sigma^2 = 4)$  to evaluate the integral.
- (b) Estimate the error  $\sigma_I$  in the integral without resorting to multiple runs.

**Points:** (a) 20, (b) 10

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<sup>2</sup>To sample  $x'$  from a normal distribution with mean  $\mu$  and variance  $\sigma^2$ , set  $x' = \sigma x + \mu$ , where  $x$  is sampled from a unit normal in Q1. You may use a built-in Gaussian random number generator as a fallback.

### 3. Bayesian Inference

Suppose the number of major hurricanes  $k$  in the North Atlantic per year follows a Poisson distribution,  $p(k|\lambda) = (\lambda^k e^{-\lambda})/k!$ , with an average rate of  $\lambda$  per year.

The number of such hurricanes over a  $n = 5$  year period between 2017-2021 was  $k_1 = 6$ ,  $k_2 = 2$ ,  $k_3 = 3$ ,  $k_4 = 6$ ,  $k_5 = 4$ . If we assume that these observations are independent of each other, then the combined likelihood is given by,

$$p(\mathbf{k}|\lambda) = \prod_{i=1}^5 p(k_i|\lambda), \quad (5)$$

where  $\mathbf{k} = [k_1, k_2, k_3, k_4, k_5]$  is the set of  $n = 5$  observations. Our goal is to estimate the posterior distribution  $p(\lambda|\mathbf{k})$  from these observations by using Bayesian inference:

$$p(\lambda|\mathbf{k}) \propto p(\mathbf{k}|\lambda) p(\lambda). \quad (6)$$

Suppose we assume that the prior distribution  $p(\lambda)$  is a Gamma distribution,

$$p(\lambda) = \frac{b^a}{\Gamma(a)} \lambda^{a-1} e^{-b\lambda}$$

with parameters  $a = 5$  and  $b = 0.5$ , and  $\Gamma(\cdot)$  is the Gamma function. Since Gamma and Poisson are conjugate distributions, it can be shown that the posterior distribution,

$$p(\lambda|\mathbf{k}) = \frac{\beta^\alpha}{\Gamma(\alpha)} \lambda^{\alpha-1} e^{-\beta\lambda}$$

is also a Gamma distribution with parameters  $\alpha = a + \sum_i k_i$  and  $\beta = n + b = 5 + b$ .

- (a) Plot the prior and posterior distributions, and comment on difference between them.<sup>3</sup>
- (b) Suppose the prior  $p(\lambda)$  distribution was uniform  $U[0, 20]$ . Describe a step-by-step approach for estimating the posterior distribution. You are not required to implement this part, but if you do, I will give you 10 bonus points.

**Points:** (a) 15, (b) 10 (+10 bonus).

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<sup>3</sup>**Hint:** You are not required to do any Monte Carlo sampling for this problem.