

## 5 Monte Carlo (AJ Chapter 8)

The (\*) questions from sheets 5 to 9 form homework 2.

1. [The Monte Carlo method for calculating  $\int h(x)f(x)dx$  where  $f$  is a p.d.f is to sample  $X_1, \dots, X_n \sim f$  then calculate  $\frac{1}{n} \sum_{i=1}^n h(X_i)$ .]
  - (a) Estimate the absolute value of a normal  $N(0, 1)$  random variable. Estimate how accurate your answer is.
  - (b) How does  $\int_{[0,1]^n} \int_{[0,1]^n} \|x - y\|_p dx dy$  behave when  $n$  is large (and  $p \in \{1, 2\}$ , say). [This is related to *the curse of dimensionality*.]
  - (c) Suppose some statistic  $X_n$  is measured annually ( $n = 2014, 2015, \dots$ ; the number of accidents per year or something). Suppose the  $X_n$  are independent samples from the  $\text{Poisson}(\lambda)$  distribution. How large does  $\lambda$  have to be for the percentage change (this year compared to last year) to be typically less than 10%.
2. (\*) [Inversion sampling: The c.d.f. for a random variable  $X$  is

$$F(x) = \mathbb{P}(X \leq x).$$

The generalized inverse c.d.f. is

$$F^{-1}(u) = \inf\{x : F(x) \geq u\}.$$

If  $U \sim \text{Uniform}(0, 1)$  then  $F^{-1}(U)$  has c.d.f.  $F$ . ]

Use the inversion method, and the R function `runif`, to sample from the geometric distribution on  $1, 2, \dots$  with mean  $1/p$ ,

$$\mathbb{P}(X = k) = (1 - p)^{k-1}p, \quad k = 1, 2, \dots$$

3. (\*) Let  $U$  denote the set

$$\{(x_1, x_2, x_3) \in (0, 1)^3 : \sqrt{x_1} + \sqrt[3]{x_2} + \sqrt[4]{x_3} < 1\}.$$

We can estimate the volume of  $U$  using the rectangle method, applied to the indicator function of  $U$

```
n=100; i=(2*(1:n)-1)/(2*n)
a=0
for (j1 in i) {
  for (j2 in i) {
    for (j3 in i) {
      if (j1**(1/2)+j2**(1/3)+j3**(1/4)<1) a=a+1
    }
  }
}
print(a/n**3)
[1] 0.000724
```

- (a) Sampling from the uniform distribution on  $[0, 1]^2$  or  $[0, 1]^3$ , use Monte Carlo integration to estimate the volume of  $U$ . Estimate the accuracy of your answers.
- (b) [Rejection sampling: Given a two distributions  $f$  and  $g$  such that  $f(x) \leq cg(x)$  for all  $x$ , to sample from the  $f$  distribution:
- sample from the  $g$  distribution
  - accept the sample with probability  $f(x)/(cg(x))$ . If you reject, try again.]

Use rejection sampling to sample  $(X_1, X_2, X_3)$  from the uniform distribution on  $U$ . Estimate the expected values of  $X_1, X_2$  and  $X_3$ , and the accuracy of your estimates.

- (c) [Importance sampling: Estimate  $\int h(x)f(x)dx$  by sampling  $Y_1, \dots, Y_n \sim g$  and then calculate either

$$\frac{1}{n} \sum_{i=1}^n h(Y_i)f(Y_i)/g(Y_i)$$

or (if you only know  $f$  up to a multiplicative constant)

$$\sum_{i=1}^n [h(Y_i)f(Y_i)/g(Y_i)] / \sum_{i=1}^n [f(Y_i)/g(Y_i)]$$

]

Sample  $Y_1, Y_2, Y_3$  independently from the exponential distribution, with mean values as computed in part (b). Use  $Y_1, Y_2, Y_3$  and importance sampling to estimate the expected values of  $X_1, X_2, X_3$  when  $(X_1, X_2, X_3)$  is uniformly distributed on  $U$ . Estimate the accuracy of your estimates.

4. [Sequential importance sampling with rejection control: SIS is an incremental version of importance sampling. It can be combined with rejection sampling in a variety of clever ways to save you from exploring parts of the sample space with low weights.
- (a) Let  $\pi = \pi(x_1, \dots, x_d) = \pi(x_{1:d})$  denote a probability density function for  $X = (X_1, \dots, X_d) = X_{1:d} \in \mathbb{R}^d$ .
- (b) For  $i = 1, \dots, d$ , let  $\pi_i = \pi_i(x_{1:i})$  denote positive functions on  $\mathbb{R}^i$  such that  $\pi \propto \pi_d$ .
- (c) For  $i = 1, \dots, d$ , let  $g_i$  denote proposal probability density functions on  $\mathbb{R}^i$ .
- (d) Let  $c_1, \dots, c_d$  denote threshold values.

Taking the threshold values equal to zero gives ordinary importance sampling with respect to the distribution  $g_1(x_1)g_2(x_2 | x_1) \dots g_d(x_d | x_{1:d-1})$ . In that case,  $\bar{w}_i \equiv w_i$  for all  $i$ .

Step 1:

Sample  $x_1 \sim g_1(x_1)$ .

Let  $w_1 = \pi_1(x_1)/g_1(x_1)$ .

Accept  $x_1$  with probability  $\min\{1, w_1/c_1\}$ . Let  $\bar{w}_1 = \max\{w_1, c_1\}$ .

Step  $t + 1$ : ( $t = 1, \dots, d - 1$ )

Sample  $x_{t+1} \sim g_t(x_{t+1} | x_{1:t})$ .

Let  $w_{t+1} = \bar{w}_t \pi_{t+1}(x_{1:t+1}) / [g_{t+1}(x_{t+1} | x_{1:t}) \pi_t(x_{1:t})]$ .

Accept  $x_{t+1}$  with probability  $\min\{1, w_{t+1}/c_{t+1}\}$ .

Let  $\bar{w}_{t+1} = \max\{w_{t+1}, c_{t+1}\}$ .

If  $Y$  is a sample from the SIS procedure then

$$\mathbb{E}_{X \sim \pi}[h(X)] = \frac{\mathbb{E}[h(Y)\bar{w}_d]}{\mathbb{E}[\bar{w}_d]}.$$

]

Suppose that the logarithm of the size of a population of fish at times  $t = 1, \dots, d$  is given by the vector  $X = (X_1, \dots, X_d)$ . Assume that  $X$  is an auto-regressive process ( $|\lambda| < 1$ ),

$$X_1 = Z_1, \quad Z_1 \sim \text{Normal}(\mu, \sigma^2/(1 - \lambda^2))$$

$$X_t = Z_t + \lambda X_{t-1}, \quad Z_t \sim \text{Normal}((1 - \lambda)\mu, \sigma^2), \quad t = 2, \dots, d$$

The population  $X$  cannot be observed directly. Instead we observe the number of fish  $Y$  caught in nets

$$Y_t \sim \text{Poisson}(\exp(X_t)/100), \quad t = 1, \dots, d.$$

Let  $f$  denote the joint distribution of  $X$  and  $Y$

$$f(x, y) = f(x)f(y | x).$$

We can use SIS to to sample from the posterior distribution

$$\pi(x) := f(x | y) \propto f(x)f(y | x).$$

For  $t = 1, \dots, d$  let

$$\pi_t(x_{1:t}) = f(x_{1:t})f(y_{1:t} | x_{1:t}), \quad g_t(x_{1:t}) = f(x_{1:t}).$$

The importance weights are given by

$$w_1 = \frac{\pi_1(x_1)}{g_1(x_1)} = f(y_1 | x_1).$$

$$w_{t+1} = \bar{w}_t \frac{\pi_{t+1}(x_{1:t+1})}{\pi_t(x_{1:t})g_{t+1}(x_{t+1} | x_{1:t})} = \bar{w}_t f(y_{t+1} | x_{t+1})$$

```

ag_process<-function(mu,sigma,lambda,n) {
  x=c()
  x[1]=rnorm(1,mu,sigma/sqrt(1-lambda**2))
  for (i in 2:n)
    x[i]=x[i-1]*lambda + rnorm(1,mu*(1-lambda),sigma)
  x
}
#Generate observations Y
x=ag_process(10,0.3,0.75,10) #Pretend you cannot see this!
y=rpois(10,exp(x)/100)

```

- (a) Use SIS (without rejection control) to sample  $x$  conditional on  $y$ . Produce an empirical plot of the distribution of the log weights.
- (b) Implement SIS with rejection control.
- (c) How can part (b) be made more efficient? [Rejection Control and Sequential Importance Sampling. Jun S. Liu; Rong Chen; Wing Hung Wong.[http://www.people.fas.harvard.edu/~junliu/TechRept/98folder/liu98\\_s.pdf](http://www.people.fas.harvard.edu/~junliu/TechRept/98folder/liu98_s.pdf)]