

## Stochastic Modeling

### MCMC for estimating the cardinality of a set

The objective is to study Monte Carlo methods for estimating the cardinality  $N(d)$  of the set  $A_d = \{(x_1, \dots, x_d) \in \mathbb{Z}^d \cap [0, 5]^d, x_1 + \dots + x_d = 5\}$ . The first method is a crude Monte Carlo method (in this case equivalent to a reject method), the second one is a more advanced MCMC method. The aim is to show the non efficiency of the first one when  $d$  is too big and the relative efficiency of the second one.

**Remark :** It is a difficult problem and both methods fail when the dimension  $d$  is too big.

### Preliminaries

Calculate  $N(1)$ ,  $N(2)$  and  $N(3)$ ? It starts to be tedious, is not it? Easier one is smart and good in combinatorics and one finds the exact calculation  $N(d) = \binom{d+4}{d-1}$ , or one uses MC methods.

### Crude Monte Carlo method

1. Give the a.s. limit of the sequence  $\frac{1}{n} \sum_{i=1}^n 1_{Y_i \in A_d}$ ,  $n \rightarrow \infty$ , for iid  $Y_i$ s uniformly distributed over  $\mathbb{Z}^d \cap [0, 5]^d$ ?
2. Write a function `MCsimple` with arguments the dimension  $d$  and the sample size  $n$  providing an estimation  $\hat{N}_1(d)$  of  $N(d)$  based on the crude Monte Carlo method over  $n$  samples. Test the function to estimate  $N(2)$  and  $N(3)$ .
3. Illustrate the randomness of the algorithm thanks to boxplots for  $d = 3, 5$  et  $10$  pour  $n = 1000$ . What is happening when the dimension is too big? Print the function  $d \mapsto 6^{-d} \binom{d+4}{d-1}$ . Is it reasonable to use the crude MC method for estimating  $N(10)$ ?

### MCMC method

3. What is the code below producing?

```
walkd <- function(d){
  walk <- rep(0,d)
  u <- runif(1)
  coord1 <- ceiling(d*u)
  v <- runif(1)
  coord2 <- ceiling(d*v)
  if (coord1 != coord2) {
    walk[coord1] <- 1
    walk[coord2] <- -1
  }
  return(walk)
}
```

Use it to create a random walk on  $\mathbb{Z}^d$  on the set  $\{(x_1, \dots, x_d) \in \mathbb{Z}^d, x_1 + \dots + x_d = 5\}$  starting from  $(5, 0, \dots, 0)$  and represent a trajectory of length 1000 for  $d = 2$ .

4. Modify the previous code in order that the chain stays on the boundary of  $[0, 5]^d$  instead of exiting  $[0, 5]^d$ . The chain restricted in  $[0, 5]^d$  is denoted  $X$ .
5. Check that the Markov chain is reversible. Show the existence and unicity of the stationary measure of  $X$ . Determine the stationary measure which is also the invariant measure. What are the ergodic properties of  $X$ ?
6. We want to estimate  $N(4)$  thanks to  $X$ . Give the cardinality of  $B_4 \subset A_4$  defined as

$$B_4 := \{(x_1, x_2, x_3, 0) \in \mathbb{N}^4, x_1 + \dots + x_3 = 5\}.$$

Give the a.s. limit of  $\frac{1}{n} \sum_{i=1}^n \mathbf{1}_{X_i \in B_4}$ ,  $n \rightarrow \infty$ ? Estimate  $N(4)$  (choose  $n = 1000$ ).

7. Define recursively

$$B_d := \{(x_1, \dots, x_{d-1}, 0) \in \mathbb{N}^d, x_1 + \dots + x_{d-1} = 5\}, \quad d \geq 1.$$

What is the cardinality of  $B_d$ ? Deduce a method to estimate recursively  $N(k)$  for  $k = 4, \dots, d$ . Let  $\hat{N}_2(d)$  the corresponding estimation of  $N(d)$ .

8. Compare the estimators  $\hat{N}_1(d)$  and  $\hat{N}_2(d)$  for increasing values of  $d$  thanks to boxplots on which you represent the true value of  $N(d)$ .