

Computer simulations of Markov Chains

All major programming languages have a random numbers generator, producing sequences of i.i.d. numbers

$$U_0, U_1, U_2, \dots \sim U(0, 1)$$

(Actually they are pseudo-random numbers generators \Rightarrow possible problem, but we disregarded it)

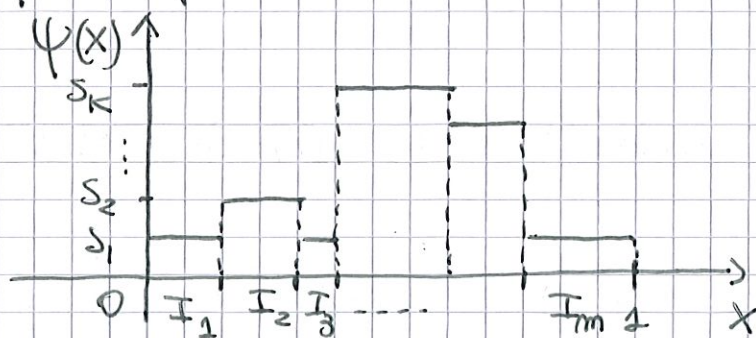
Starting from this, we want to simulate a Markov Chain $\{X_0, X_1, X_2, \dots\}$ with given state space $S = \{s_1, s_2, \dots, s_k\}$, initial distrib. $\mu^{(0)}$ and transition matrix P (constant \Rightarrow homog. MC).

Main ingredients: initiation function and update function.

The initiation function $\psi: [0, 1] \rightarrow S$ is a function that we use to generate the starting value X_0 .

Assumptions:

(i) ψ is piecewise constant:



$[0, 1]$ partitioned into intervals $\{I_1, I_2, \dots, I_m\}$ where ψ is constant

(ii) for each $s \in S$, the total length of the intervals on which $\psi(x) = s$ equals $\mu^{(0)}(s) = \mathbb{P}(X_0 = s)$

$$\sum_{\{j: \psi(x) = s \text{ on } I_j\}} l(I_j) = \mu^{(0)}(s)$$

$$l(I_j) = \text{length } I_j$$

A method to build an initiation function which satisfies (i) and (ii) is the following (inverse transform method):

Let $S = \{s_1, \dots, s_k\}$ $\mu^{(0)} = (\mu^{(0)}(s_1), \dots, \mu^{(0)}(s_k))$
we can set:

$$\psi(x) = \begin{cases} s_1 & \text{for } x \in [0, \mu^{(0)}(s_1)] = I_1 \\ s_2 & \text{for } x \in [\mu^{(0)}(s_1), \mu^{(0)}(s_1) + \mu^{(0)}(s_2)] = I_2 \\ \vdots & \\ s_i & \text{for } x \in \left[\sum_{j=1}^{i-1} \mu^{(0)}(s_j), \sum_{j=1}^i \mu^{(0)}(s_j) \right] = I_i \\ \vdots & \\ s_k & \text{for } x \in \left[\sum_{j=1}^{k-1} \mu^{(0)}(s_j), 1 \right] = I_k \end{cases}$$

why it works?

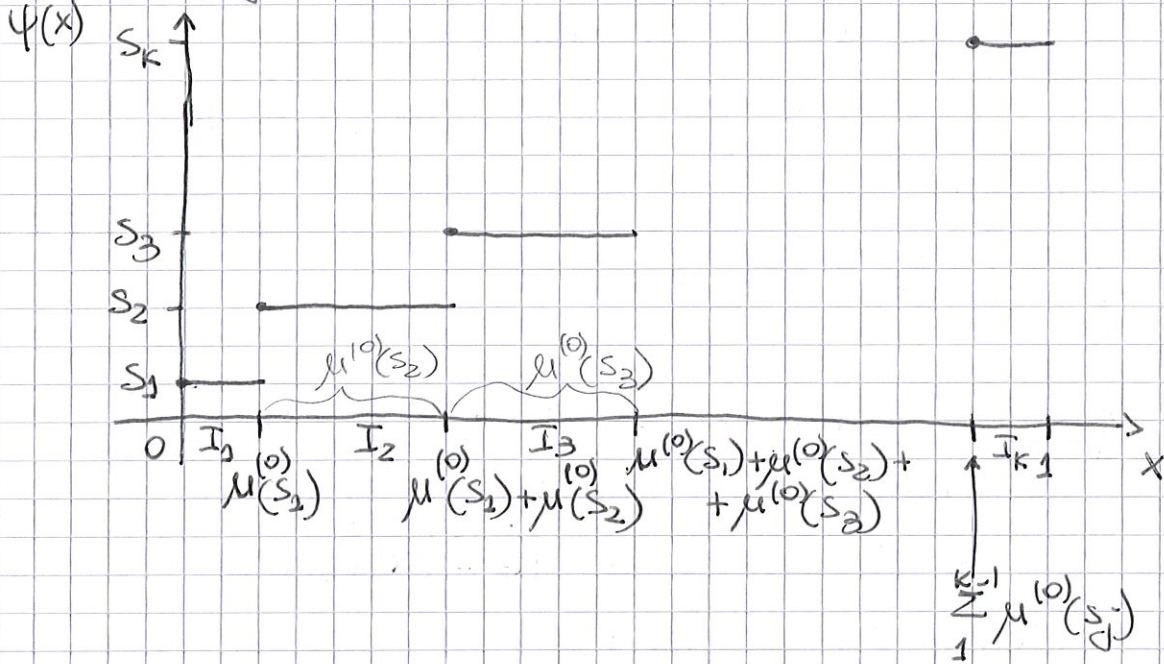
(i) is clearly satisfied ($\psi(x)$ is piecewise constant)

Let's verify that (ii) is also satisfied:

For each s_i : $\psi(x) = s_i$ only on the interval $\left[\sum_{j=1}^{i-1} \mu^{(0)}(s_j), \sum_{j=1}^i \mu^{(0)}(s_j) \right)$, whose length is

$$\sum_{j=1}^i \mu^{(0)}(s_j) - \sum_{j=1}^{i-1} \mu^{(0)}(s_j) = \mu^{(0)}(s_i) \quad \text{ok!}$$

graphically:



remember that since $\mu^{(0)}$ is a distribution,
 $\sum_{i=1}^k \mu^{(0)}(s_i) = 1 \Rightarrow$ we are partitioning correctly
 the interval $[0, 1]$

Thus we can simulate X_0 using the initiation function $\psi(x)$ as follows:

Algorithm to simulate X_0

- Generate a random number $U \sim U(0, 1)$
- Identify the interval I_ℓ in which U is fallen and set $X_0 = s_\ell$

Usually b) can be performed with a "while
code":

b) counter = 0

score = 0

while $U > \text{score}$ {

counter = counter + 1

score = score + $\mu^{(0)}(S_{\text{counter}})$

}

output: $X_0 = S_{\text{counter}}$

The update function $\phi: S \times [0, 1] \rightarrow S$
is the distribution by which we generate X_{m+1}
from $X_m \Rightarrow$ iterating it we generate the
entire chain (X_0, X_1, \dots) .

$\phi(s, x)$ takes as input a state $s \in S$
and a number $x \in [0, 1]$ (a probability)
and produces another state $s' \in S$ as
output.

Assumptions:

(i) for any fixed s_i , $\phi(s_i, x)$, as function of x ,
- is piecewise constant

(ii) for each fixed $s_i, s_j \in S$, the total length of
the intervals on which $\phi(s_i, x) = s_j$ equals
 $P_{ij} = \mathbb{P}(X_{m+1} = s_j \mid X_m = s_i)$ (element ij of P)

Remember that "rows of P sum to one"

$$P = \begin{bmatrix} \dots & & \\ P_{i1} & \dots & P_{ik} \\ \dots & & \end{bmatrix} \quad \left. \begin{array}{l} i\text{-th row: can be seen} \\ \text{as a distribution} \end{array} \right\}$$

thus we may define our update function ϕ satisfying (i) and (ii) similarly as before, replacing $x^{(0)}$ with the i -th row of P :

$$\phi(s_i, x) = \begin{cases} s_1 & \text{for } x \in [0, P_{i1}] = I_1 \\ s_2 & \text{for } x \in [P_{i1}, P_{i1} + P_{i2}] = I_2 \\ \vdots & \vdots \\ s_j & \text{for } x \in \left[\sum_{l=1}^{j-1} P_{il}, \sum_{l=1}^j P_{il} \right) = I_j \\ \vdots & \vdots \\ s_k & \text{for } x \in \left[\sum_{l=1}^{k-1} P_{il}, 1 \right] = I_k \end{cases}$$

It can be proven as before that this definition is correct and satisfies the assumptions.

Algorithm to simulate X_{m+1} given X_m

- Generate a random number $U \sim U(0, 1)$
- consider $\phi(X_m, x)$: check in which interval I_ℓ U is fallen and set $X_{m+1} = s_\ell$

General algorithm to simulate the MC

$$\begin{aligned} X_0 &= \psi(U_0) \\ X_1 &= \phi(X_0, U_1) \\ X_2 &= \phi(X_1, U_2) \\ &\vdots \end{aligned} \quad U_i \sim U(0, 1) \text{ independent}$$

Example - The Gothenburg weather

Remember that in our example we had

$S = \{s_1, s_2\}$ $s_1 = \text{"rain"}$ $s_2 = \text{"sunshine"}$
and transition matrix

$$P = \begin{bmatrix} 0.75 & 0.25 \\ 0.25 & 0.75 \end{bmatrix}$$

Suppose that we start the MC in a rainy day $\Rightarrow X_0 = s_1$ and $\mu^{(0)} = (1, 0)$

\Rightarrow Initiation function: $\psi(x) = s_1 \quad \forall x$

Update function:

$$\phi(s_1, x) = \begin{cases} s_1 & \text{for } x \in [0, 0.75) \\ s_2 & \text{for } x \in [0.75, 1] \end{cases}$$

$$\phi(s_2, x) = \begin{cases} s_1 & \text{for } x \in [0, 0.25) \\ s_2 & \text{for } x \in [0.25, 1] \end{cases} \quad (*)$$

Simulation of an inhomogeneous Markov Chain

Extension (quite simple) of previous method:

let (X_0, X_1, \dots) be an inhomogeneous MC with

state space $S = \{s_1, s_2, \dots, s_k\}$

initial distribution $\mu^{(0)}$

transition matrices $P^{(0)}, P^{(1)}, P^{(2)}, \dots$

We want to simulate this MC.

We can obtain the initiation function ψ and the starting value X_0 exactly like in the homogeneous case.

For the updating we need a sequence of updating functions $\phi^{(1)}, \phi^{(2)}, \dots$ which are computed as before, but using at each time step the corresponding transition matrix:

$$\phi^{(m)}(s_i, x) = \begin{cases} s_1 & \text{for } x \in [0, P_{i1}^{(m)}) \\ s_2 & \text{for } x \in [P_{i2}^{(m)}, P_{i2}^{(m)} + P_{i2}^{(m)}) \\ \vdots \\ s_j & \text{for } x \in [\sum_{l=1}^{j-1} P_{il}^{(m)}, \sum_{l=1}^j P_{il}^{(m)}) \\ \vdots \\ s_k & \text{for } x \in [\sum_{l=1}^{k-1} P_{il}^{(m)}, 1] \end{cases}$$

The inhomogeneous MC is simulated by setting

$$\begin{aligned} X_0 &= \psi(U_0) \\ X_1 &= \phi^{(1)}(X_0, U_1) \\ X_2 &= \phi^{(2)}(X_1, U_2) \\ X_3 &= \phi^{(3)}(X_2, U_3) \\ &\vdots \end{aligned}$$

Exercise 3.2 p. 31 Häggström: The choice of the update function is not necessarily unique!

Consider the example of the Gothenburg weather. Show that we get another valid update function if we replace (*) with

$$(**) \quad \phi(s_2, x) = \begin{cases} s_2 & \text{for } x \in [0, 0.75) \\ s_1 & \text{for } x \in [0.75, 1] \end{cases}$$

Solution:

Clearly $\phi(s_i, x)$ is piecewise constant for each s_i . Let's check if (ii) is valid:

Consider s_1 :

$$\phi(s_1, x) = s_1 \quad \text{if } x \in I_1 = [0, 0.75) \quad l(I_1) = 0.75 = P_{11}$$

$$\phi(s_1, x) = s_2 \quad \text{if } x \in I_2 = [0.75, 1] \quad l(I_2) = 0.25 = P_{12}$$

ok!

Consider s_2 in (**):

$$\phi(s_2, x) = s_1 \quad \text{if } x \in I_2 = [0.75, 1] \quad l(I_2) = 0.25 = P_{22}$$

$$\phi(s_2, x) = s_2 \quad \text{if } x \in I_1 = [0, 0.75) \quad l(I_1) = 0.75 = P_{21}$$

ok!

Note that:

