

## Computer Simulations of Markov Chains

All major programming languages have a random number 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 generated  $\Rightarrow$  possible problem, but we disregarded it)

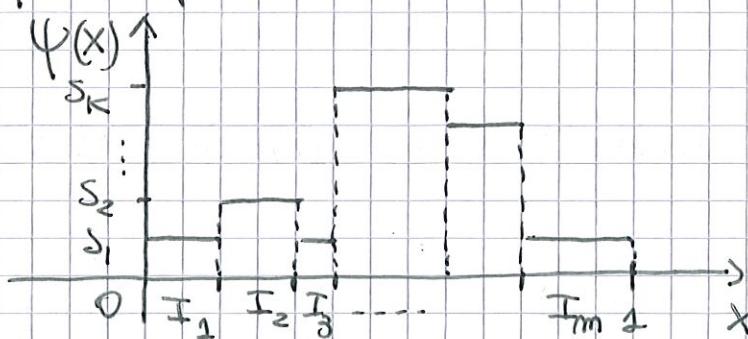
Starting from this, we want to simulate a Markov Chain  $\{X_0, X_1, X_2, \dots\}$  with given State space  $S = \{S_1, S_2, \dots, S_K\}$ , initial distib.  $\mu^{(0)}$  and transition matrix  $P$  (constant  $\Rightarrow$  homog. MC).

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)  $\psi$  is piecewise constant:



$[0,1]$  partitioned into intervals

$\{I_1, I_2, \dots, I_{m-1}\}$  where  $\psi$  is constant

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

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

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

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

$$\text{Let } S = \{s_1, \dots, s_k\} \quad \mu^{(0)} = (\mu^{(0)}(s_1), \dots, \mu^{(0)}(s_k))$$

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?

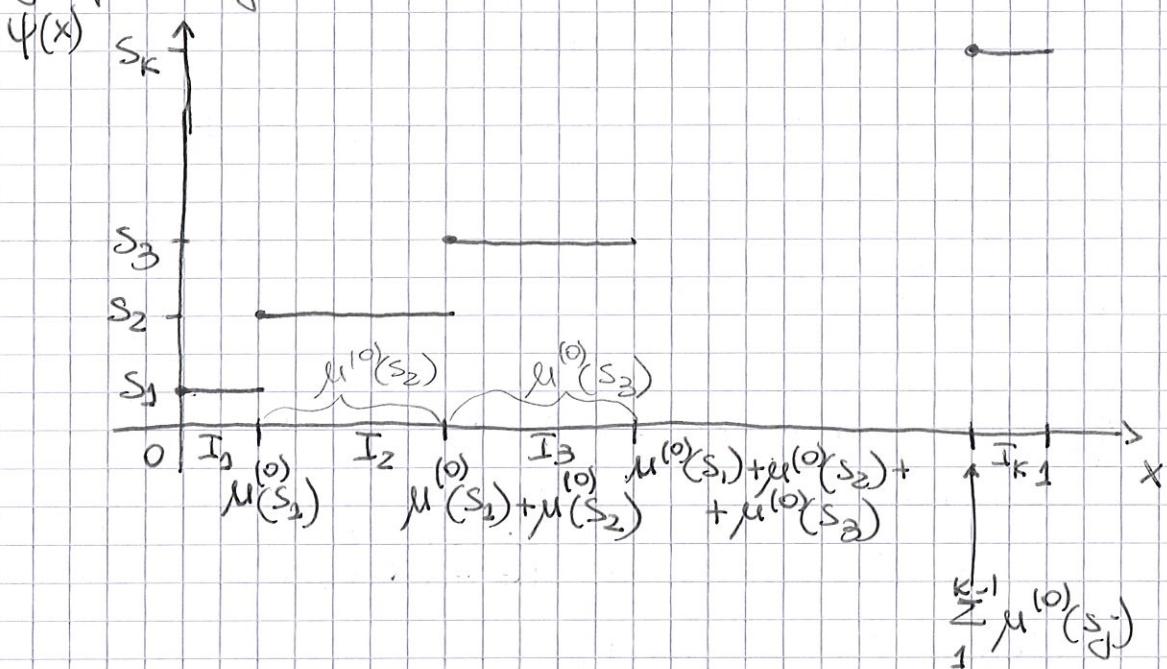
(i) is clearly satisfied ( $\psi(x)$  is piecewise constant)

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

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

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 imitation  
 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  $X_0 = s_l$

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

b) counter = 0

scall = 0

while U > scan {

counter = counter + 1

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

}

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  
 $P_{ij} = P(X_{m+1} = s_j | X_m = s_i)$  (element  $i,j$  of  $P$ )

Remember that "rows of  $P$  sum to one"

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

thus we may define our update function of satisfying  
 (i) and (ii) similarly as before, replacing  $\ell^{(t)}$  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  $X_{m+1}$  from  $X_m$

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

General algorithm to simulate the MC

$$\begin{aligned} X_0 &= \psi(U_0) \\ X_1 &= \phi(X_0, U_1) \\ X_2 &= \phi(X_1, U_2) \\ &\vdots \end{aligned}$$

$U_i \sim U(0,1)$  independent

## Example - The Gothenburg weather

Remember that in our example we had

$$S = \{S_1, S_2\} \quad S_1 = \text{"rainy"} \quad S_2 = \text{"sunshine"}$$

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 one 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 initialization function  $\psi$  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 we computed as before, but using at each time step the corresponding transition matrix:

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

The inhomogeneous MC is simulated by setting

$$X_0 = \psi(U_0)$$

$$X_1 = \phi^{(1)}(X_0, U_1)$$

$$X_2 = \phi^{(2)}(X_1, U_2)$$

$$X_3 = \phi^{(3)}(X_2, U_3)$$

$\vdots$

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:

