

# The Metropolis algorithm

The Metropolis algorithm is a general purpose algorithm, which can be applied to essentially any problem.

Let us recall our problem: **given a probability distribution  $\pi$  on a state space  $S$ , we wish to determine a transition matrix  $P$  which has  $\pi$  as equilibrium distribution.**

The Metropolis algorithm is made of two different steps.

- (i) **Step 1.** The system is currently (time  $i$ ) in state  $x_i = x$ . We propose a new configuration  $y \neq x$  of the system with a **proposal transition matrix  $P^{(0)}$** . The proposal matrix is arbitrary (it has no relation with the probability  $\pi_x$ ) and is chosen by the programmer.
- (ii) **Step 2.** Now we establish if the proposed new configuration  $y$  should be accepted or rejected. We accept it with probability  $A_{xy}$ . If the configuration  $y$  is accepted, we set  $x_{i+1} = y$ ; otherwise  $x_{i+1} = x_i$ . The acceptance matrix depends on the proposal matrix and on the probability distribution  $\pi_x$ . The matrix  $A_{xy}$  is a probability and therefore satisfies  $0 \leq A_{xy} \leq 1$ .

We now compute the Metropolis transition matrix  $P_{xy}$  (the probability of going from  $x$  to  $y$ ).

For  $y \neq x$ , the probability of going from  $x$  to  $y$  is the product of the probability of proposing  $y$  times the probability of accepting the proposed move:

$$P_{xy} = P_{xy}^{(0)} A_{xy} \quad x \neq y$$

The probability of remaining in  $x$  is obtained by using the conservation of probability

$$P_{xx} = 1 - \sum_{y \neq x} P_{xy} = P_{xx}^{(0)} + \sum_{y \neq x} P_{xy}^{(0)} (1 - A_{xy}).$$

The two terms represent the probability of proposing no change ( $P_{xx}^{(0)}$ ) and the sum of the probabilities that the proposed moves are not accepted. We used

$$1 = \sum_y P_{xy}^{(0)} = P_{xx}^{(0)} + \sum_{y \neq x} P_{xy}^{(0)}.$$

To have a valid we require that  $P$  is an **ergodic** Markov process and that the **detailed balance condition** is satisfied.

Necessary (but not sufficient) condition for  $P$  to be ergodic is that  $P^{(0)}$  is ergodic. The ergodicity of  $P$  should be verified explicitly (it is usually trivially satisfied if the system is characterized by continuous variables; more subtle is checking ergodicity for systems with discrete variables).

Now we require  $P$  to satisfy the **detailed-balance condition**  $\pi_x P_{xy} = \pi_y P_{yx}$  for any pair of states  $x, y$ . The condition becomes

$$\pi_x P_{xy}^{(0)} A_{xy} = \pi_y P_{yx}^{(0)} A_{yx}$$

- i) If  $x, y$  are such that  $P_{xy}^{(0)} = P_{yx}^{(0)} = 0$  (we never propose to go from  $x$  to  $y$  or vice versa) the condition is satisfied.
- ii) If  $x, y$  are such that  $P_{xy}^{(0)} = 0$  and  $P_{yx}^{(0)} > 0$ , we set  $A_{yx} = 0$ . If the system never goes from  $x$  to  $y$ , it should not go from  $y$  to  $x$ . Analogously, if  $x, y$  are such that  $P_{xy}^{(0)} > 0$  and  $P_{yx}^{(0)} = 0$ , we set  $A_{xy} = 0$ .
- iii) If  $x, y$  are such that  $P_{xy}^{(0)} > 0$  and  $P_{yx}^{(0)} > 0$ . The detailed-balance condition requires that

$$\frac{A_{xy}}{A_{yx}} = \frac{\pi_y P_{yx}^{(0)}}{\pi_x P_{xy}^{(0)}}$$

The right-hand side is known and we call it  $R_{xy}$ :

$$R_{xy} = \frac{\pi_y P_{yx}^{(0)}}{\pi_x P_{xy}^{(0)}}$$

which satisfies

$$R_{xy} = \frac{1}{R_{yx}}.$$

Now, the problem is:  
determine the acceptance matrix  $A_{xy}$  so that it satisfies the equation

$$\frac{A_{xy}}{A_{yx}} = R_{xy}$$

The Metropolis choice consists in taking

$$A_{xy} = \min(1, R_{xy})$$

Let us verify that this is a solution of the equation written above.  
Two cases: i)  $R_{xy} > 1$ ; ii)  $R_{xy} < 1$ .

**Case i) :**  $R_{xy} > 1$ , so that  $R_{yx} = 1/R_{xy} < 1$ . Therefore, we have

$$A_{xy} = 1 \quad A_{yx} = R_{yx} \rightarrow \frac{A_{xy}}{A_{yx}} = \frac{1}{R_{yx}} = R_{xy}$$

**Case ii) :**  $R_{xy} < 1$ , so that  $R_{yx} = 1/R_{xy} > 1$ . Therefore, we have

$$A_{xy} = R_{xy} \quad A_{yx} = 1 \rightarrow \frac{A_{xy}}{A_{yx}} = R_{xy}$$

The Metropolis choice is the optimal one. It gives the largest acceptance probability.

$$A_{xy} = R_{xy}A_{yx} \leq R_{xy}$$

because  $A_{yx} \leq 1$ . Moreover  $A_{xy} \leq 1$ . It follows the inequality

$$A_{xy} \leq \min(1, R_{xy})$$