

Figure 1: Illustration of algorithm 2

Algorithm 2 Simulator

```
1: N: number of neurons
2:  $\tilde{T}$ : time elapsed since last event in the system
3:  $T_{last}$ : absolute time of the last event in the system
4:  $T_{next}^i$ : absolute time of the nth event of neuron i
5:  $\mathcal{P}$ : poisson distribution
6:  $\mathcal{U}$ : uniform distribution
7:  $\tilde{f}$ : approximation function of function f ( $\tilde{f}(x) \geq f(x)$ )
8: repeat
9:   repeat ▷ Steps 0-1
10:     determine an interval  $[A, B[$  on which sampling
11:     compute the array of  $(\tilde{f}_i(V_i(A)))_{i \in \{1, \dots, N\}}$  and  $\sum_i \tilde{f}_i(V_i(A))$  on interval  $[A, B[$ 
12:      $\tilde{T} \sim \mathcal{P}(\lambda = \sum_i \tilde{f}_i)$ 
13:      $T_{next} \leftarrow T_{last} + \tilde{T}$ 
14:   until in good interval AND at least one  $\tilde{f}_i \neq 0$ 
15:    $i \leftarrow \text{argmin}_{i \in \{1, \dots, N\}} \left( \sum_j \tilde{f}_j(V_j(A)) * u \sim \mathcal{U}([0, 1]) < \tilde{f}_i(V_i(A)) \right)$  ▷ Step 2
16:    $T_{i,n} \leftarrow T_{next}$  ▷ Step 3
17:    $V_i(T_{i,n}) \leftarrow a + e^{-\lambda \tilde{T}}(V_i(T_{i,n-1}) - a) + \sigma \mathcal{N}(0, \frac{1 - e^{-2\lambda \tilde{T}}}{2\lambda})$ 
18:    $\mathbb{P}(\text{accepting spike of neuron } i) = \frac{f_i(V_i(T_{i,n}))}{\tilde{f}_i(V_i(A))}$  ▷ Step 4
19:   if spike is accepted then
20:     update potentials of all postsynaptic neurons
21: until an end condition is met
```

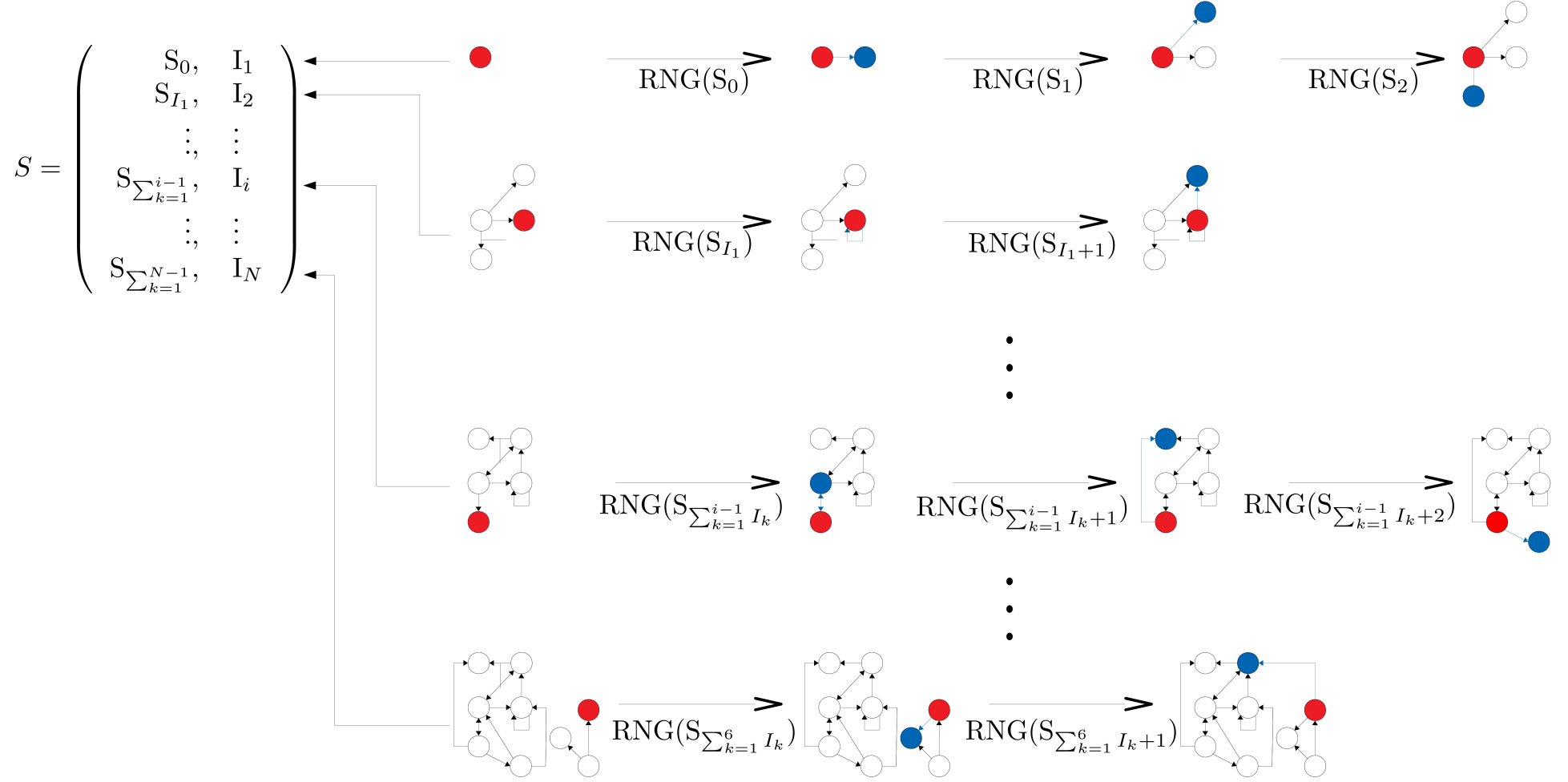


Figure 2: Construction of graph and vector of states

Algorithm 3 Generation of a matrix of children

```
1: RNG: Random Number Generator
2: p: Probability of connection between two neurons
3:  $\mathcal{M}_{i,j}$ : matrix of interaction: equals 1 for a connection, 0 otherwise
4: RNG.B(p): returns 0 or 1, bernoulli distributed with probability p.
5: function INTERACTION MATRIX(RNG, p)
6:   for i  $\leftarrow$  1 to n do
7:     for j  $\leftarrow$  1 to n do
8:        $\mathcal{M}_{i,j} \leftarrow$  RNG.B(p)
```

Algorithm 4 Generation of a vector of rng states

```
1: RNG: random Number Generator
2: p: probability of connection between two neurons
3: N: number of neurons in the system
4: S: vector of pairs (Status of the RNG, Number of children of neuron i)
5: RNG.BIN(N,p): returns a random value following the binomial distribution of parameter N and p
6: RNG( $S_i$ ): returns the index of a child for i among all other unchosen neurons
7: RNG.STATUS: returns the current internal state of the RNG
8:  $I_i$ : number of children of neuron i
9: function MAKE VECTOR(RNG, p)
10:  for i  $\leftarrow$  1 to N do
11:     $I_i \leftarrow$  RNG.BIN(N,p)
12:     $S[i] \leftarrow$  (RNG.STATUS,  $I_i$ )
13:    for j  $\leftarrow$  1 to  $I_i$  do
14:      RNG( $S_{j+\sum_{k=1}^{i-1} I_k}$ )
```

Algorithm 5 Comparison of usage between clasical method and reconstruction

```
function COMPARISON
  INTERACTION MATRIX( $RNG_1, p$ )
  MAKE VECTOR( $RNG_2, p$ )

   $\vdots$  Simulation
  Neuron i is spiking
   $\triangleright$ Using matrix graph
  for j  $\leftarrow$  1 to N do
    if  $\mathcal{M}_{i,j} = 1$  then
      Update potential of neuron j
   $\triangleright$ Using reconstruction
   $S[i] = (S_{\sum_{k=1}^{i-1} I_k}, I_i)$ 
  RNG.SETSTATE( $S[i][1]$ )
  for j  $\leftarrow$  1 to  $I_i$  do
     $Child_j \leftarrow$  RNG( $S_{j-1+\sum_{k=1}^{i-1} I_k}$ )
    Update potential of neuron  $Child_j$ 
```

	Algorithmic complexity at creation	Algorithmic complexity during usage	Memory storage
Reconstruction method	$\mathcal{O}(N^2)$	$\mathcal{O}(N)$	$\mathcal{O}(N)$
Interaction matrix	$\mathcal{O}(N^2)$	$\mathcal{O}(1)$	$\mathcal{O}(N^2)$

Table 1: Table of memory and algorithmic complexity

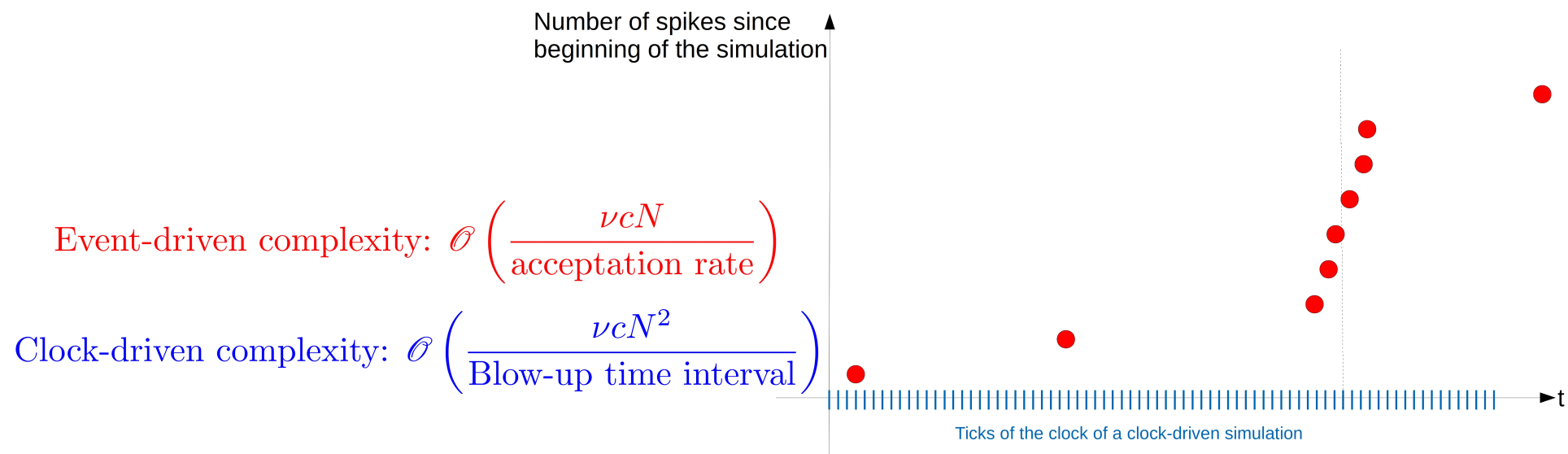


Figure 3: Complexity as number of events during a time interval