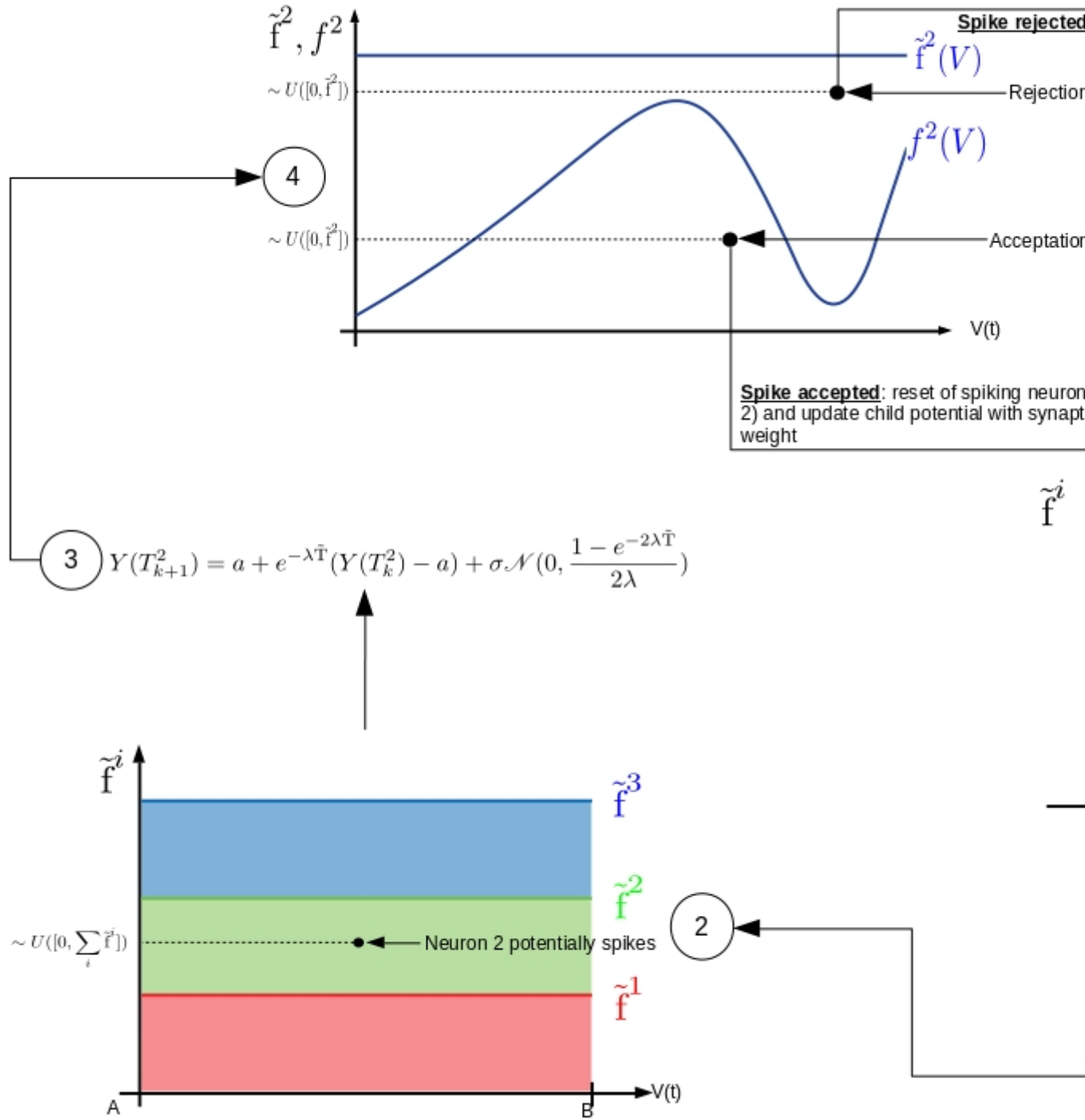


The algorithm is presented in figure ???. It explicits the several steps for simulating the network of neurons. The first step (0) is looking for a time interval, where the next event shall be looked for. The separation between two sequential events follow a Poisson distribution of parameter the sum of approximation function $T_{k+1} = T_k + \tilde{T}$, with $\mathcal{P}(\sum_i \tilde{f}^i)$.

The complexity of clock-driven algorithms is of order N^2 . Event-driven algorithms have much less complexity, of order N . But as our algorithm is based on a thinning method, the complexity is higher, but still linear of N .



(a) Pseudo code of the thinning algorithm used for simulating the system

Initialize all parameters

repeat

repeat

▷ Steps 0-1

determine an interval $[t_{n-1}, t_n]$ on which sampling

compute the array of $f_{max}(Y_{t_{n-1}})$ and $\sum f_{max}(Y_{t_{n-1}})$ on interval $[t_{n-1}, t_n]$

$t_n \sim t_{n-1} + \mathcal{E}(\sum f_{max})$

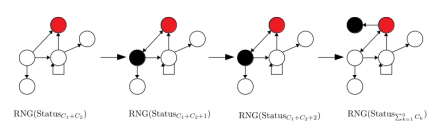
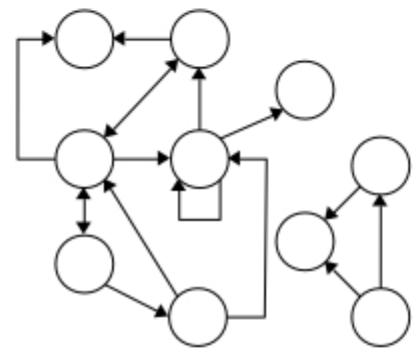
until not in good interval or all f_{max} are at 0

$$u \sim \mathcal{U}([0, 1]) \rightarrow \mathbb{P}(\text{spiking neuron is neuron } i) = \frac{f_{max}^i(Y_{t_n}^i)}{\sum_j^N f_{max}^j(Y_{t_n}^j)}$$

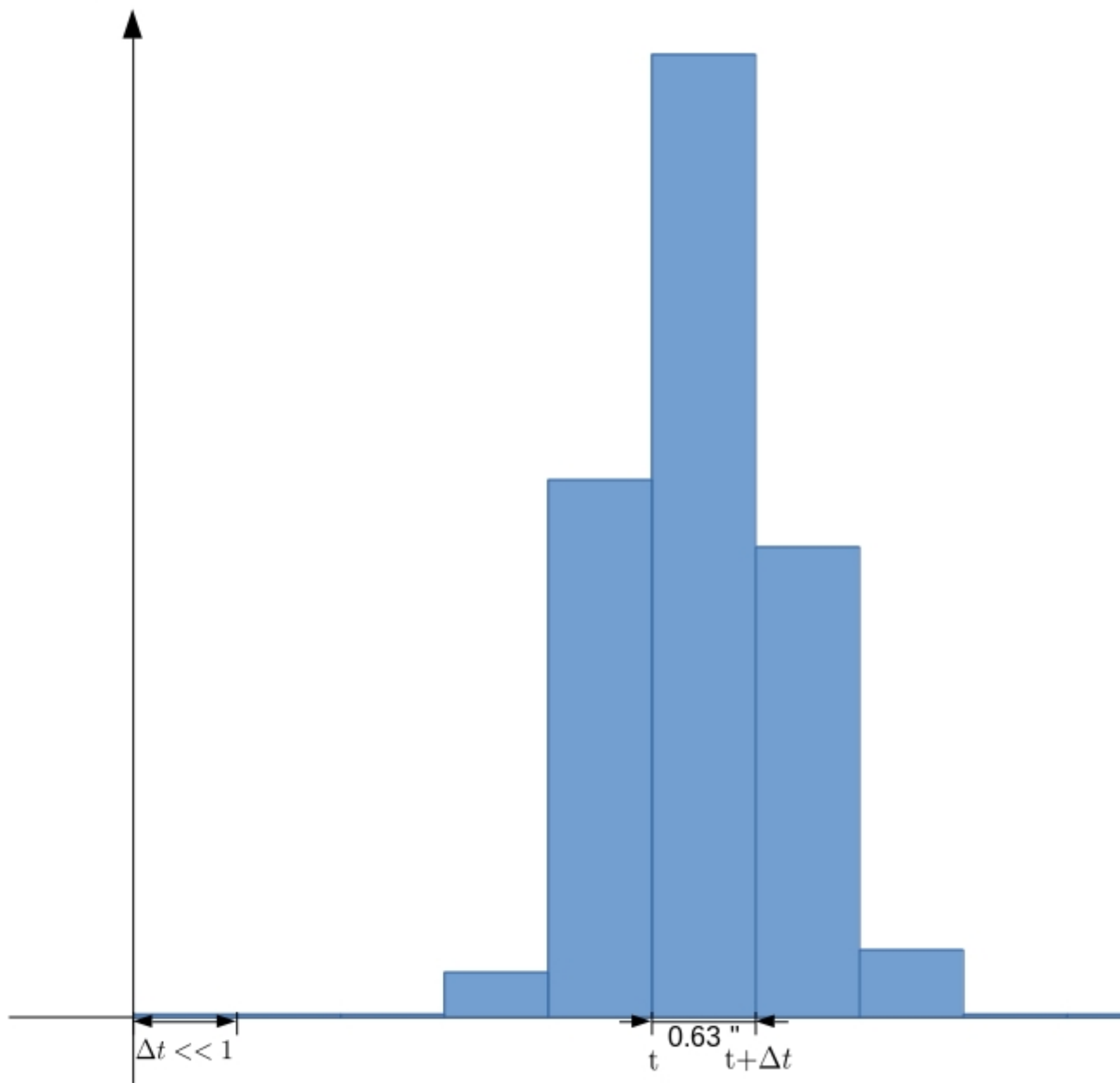
▷ Step 2

$$u \sim \mathcal{U}([0, 1]) \rightarrow \mathbb{P}(\text{accepting spike of neuron } i) = \frac{f^i(Y_{t_n}^i)}{f_{max}^i(Y_{t_n}^i)}$$

▷ Step 3



Number of spikes



(a) Complexity as number of events during a time interval

Number of spikes

