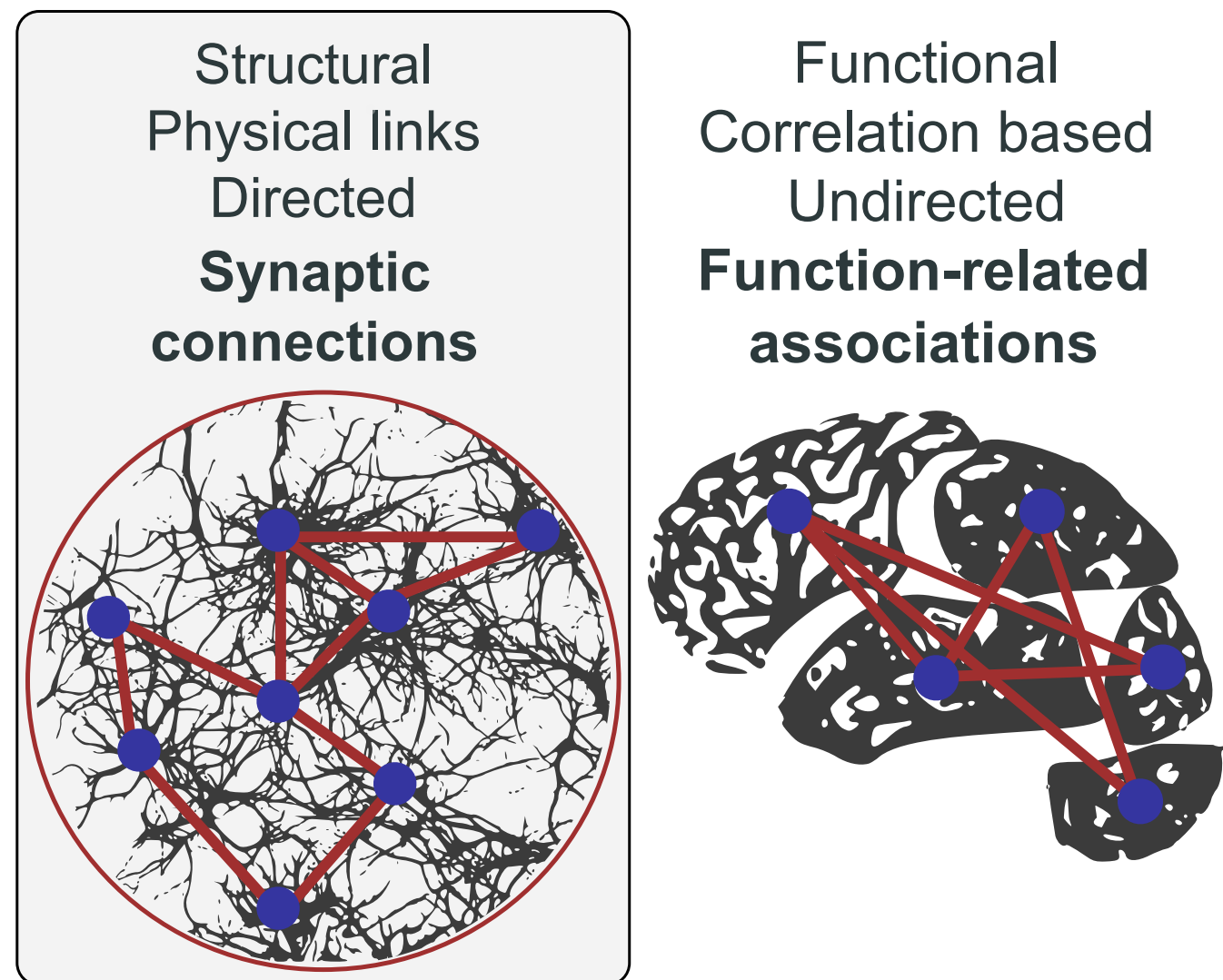
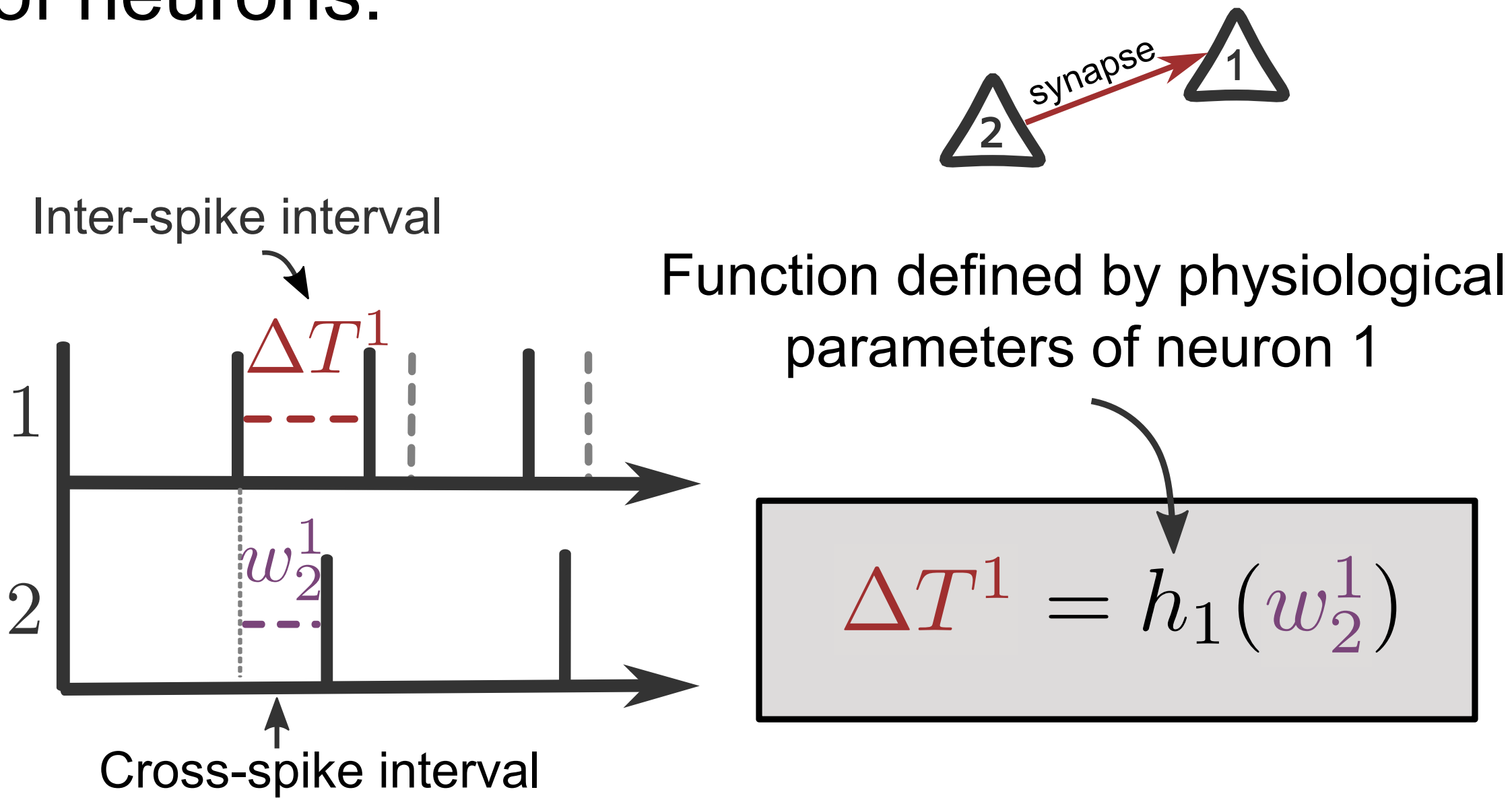


Revealing synaptic connectivity from spike train data

Types of connectivity:



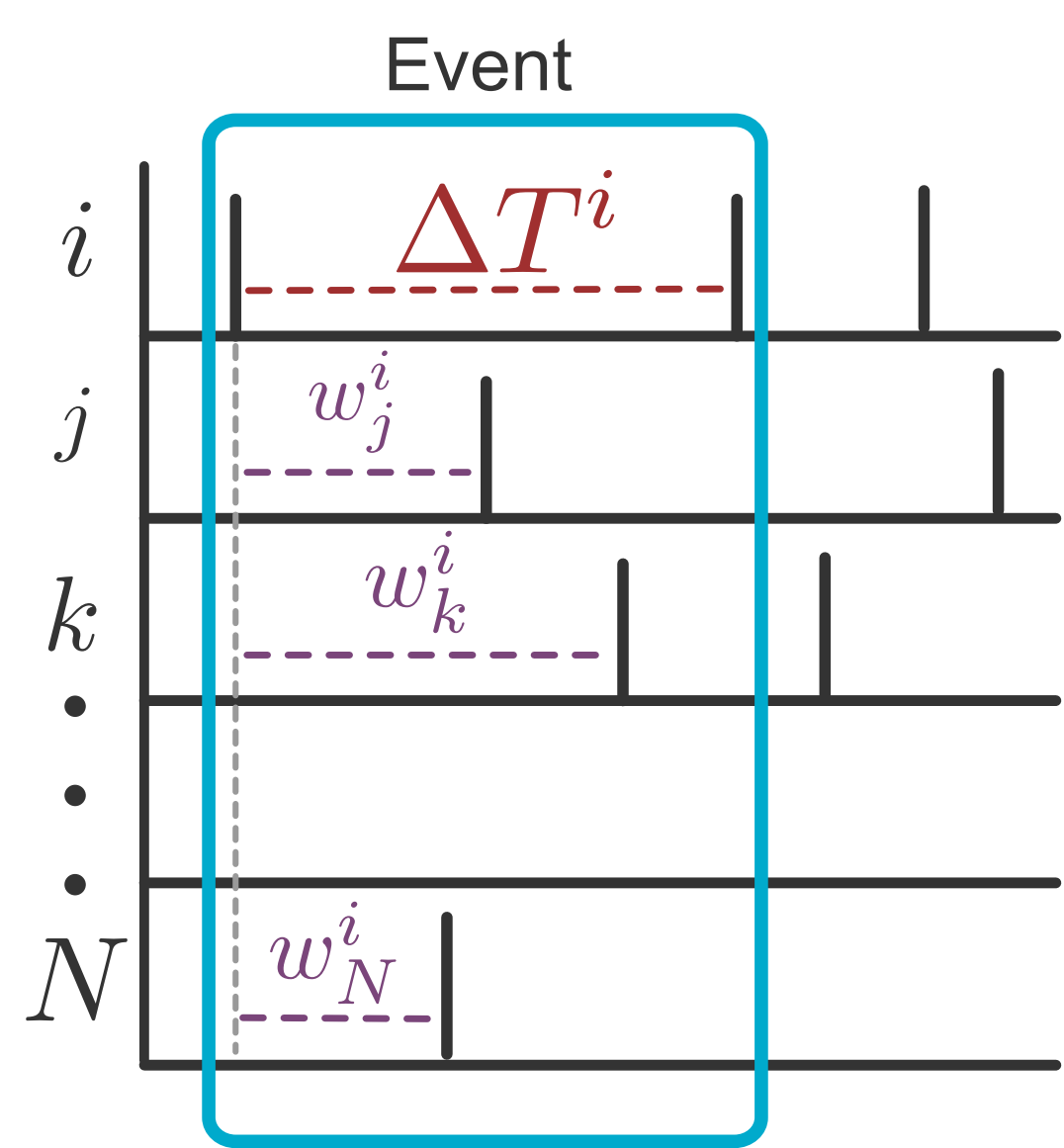
Synaptic inputs change intrinsic behaviour of neurons:



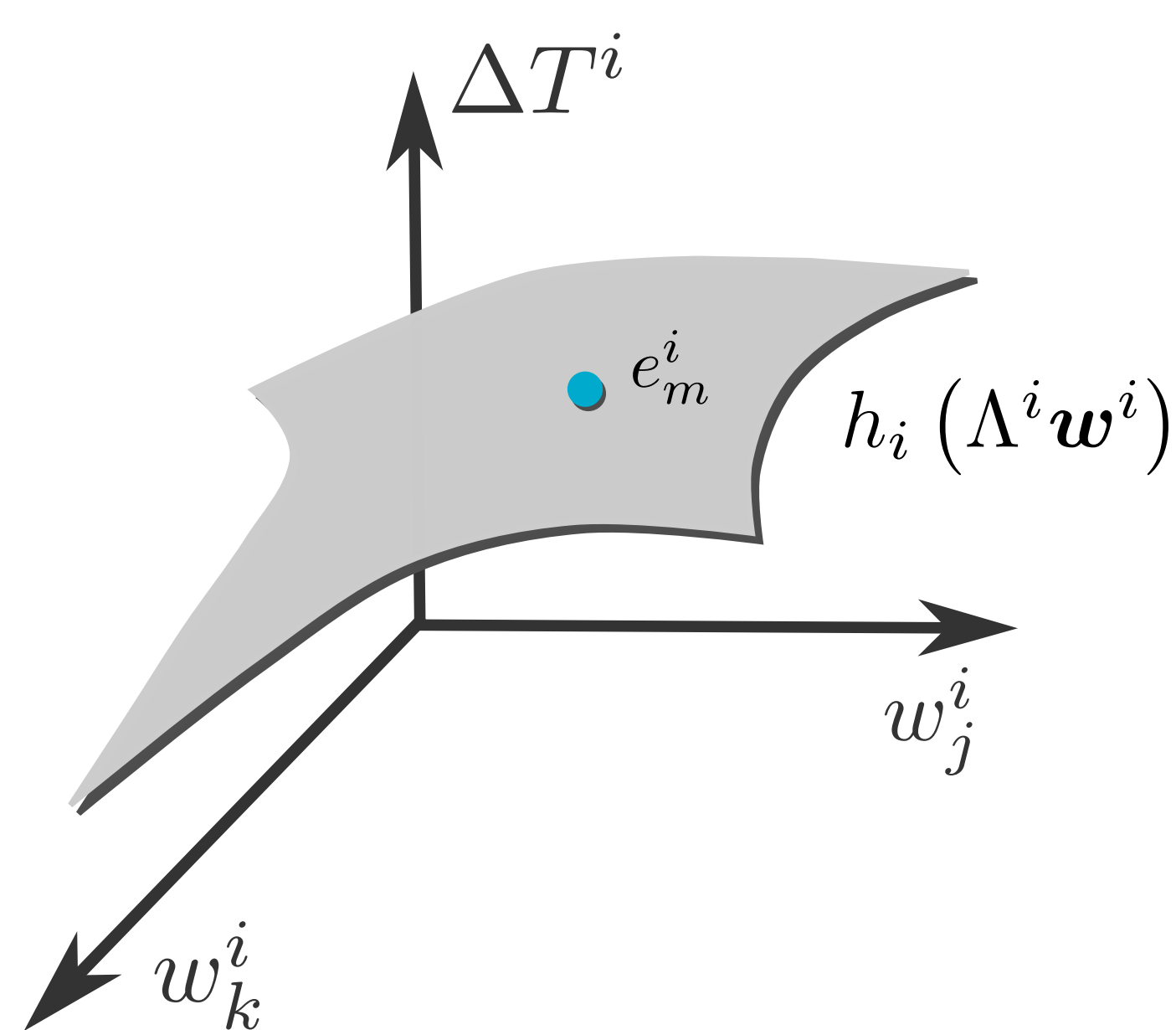
Take-home messages:

- We developed a **model-independent** approach for **revealing synaptic connections** from spike train data **only**.
- **Event space** representations do **not** require prior knowledge of physiological parameters.
- **Transmission delays** may be **estimated** from the same spike train data.

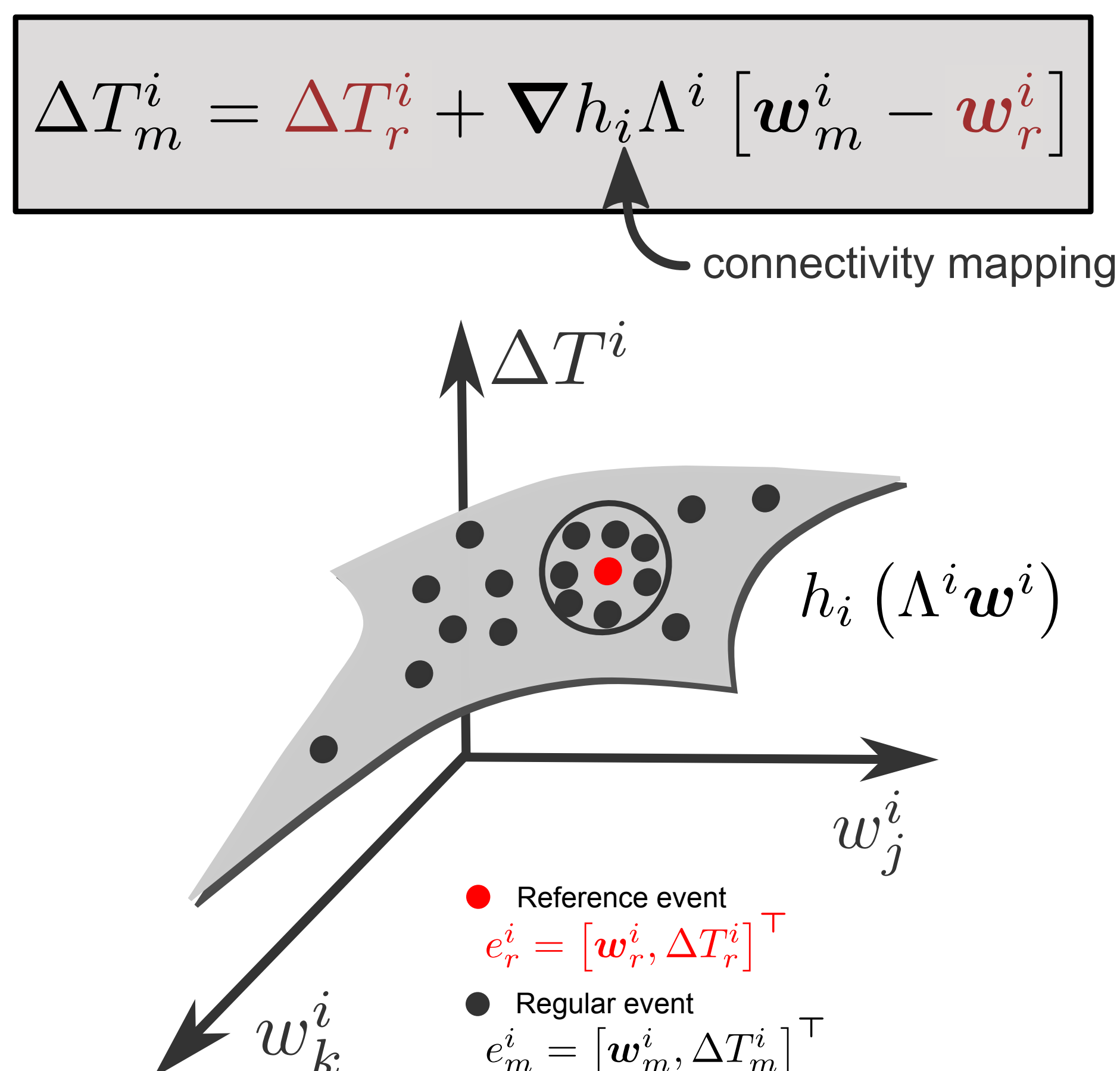
Constructing events from spike trains:



Event space representation:



Local samplings reveal synaptic connections:



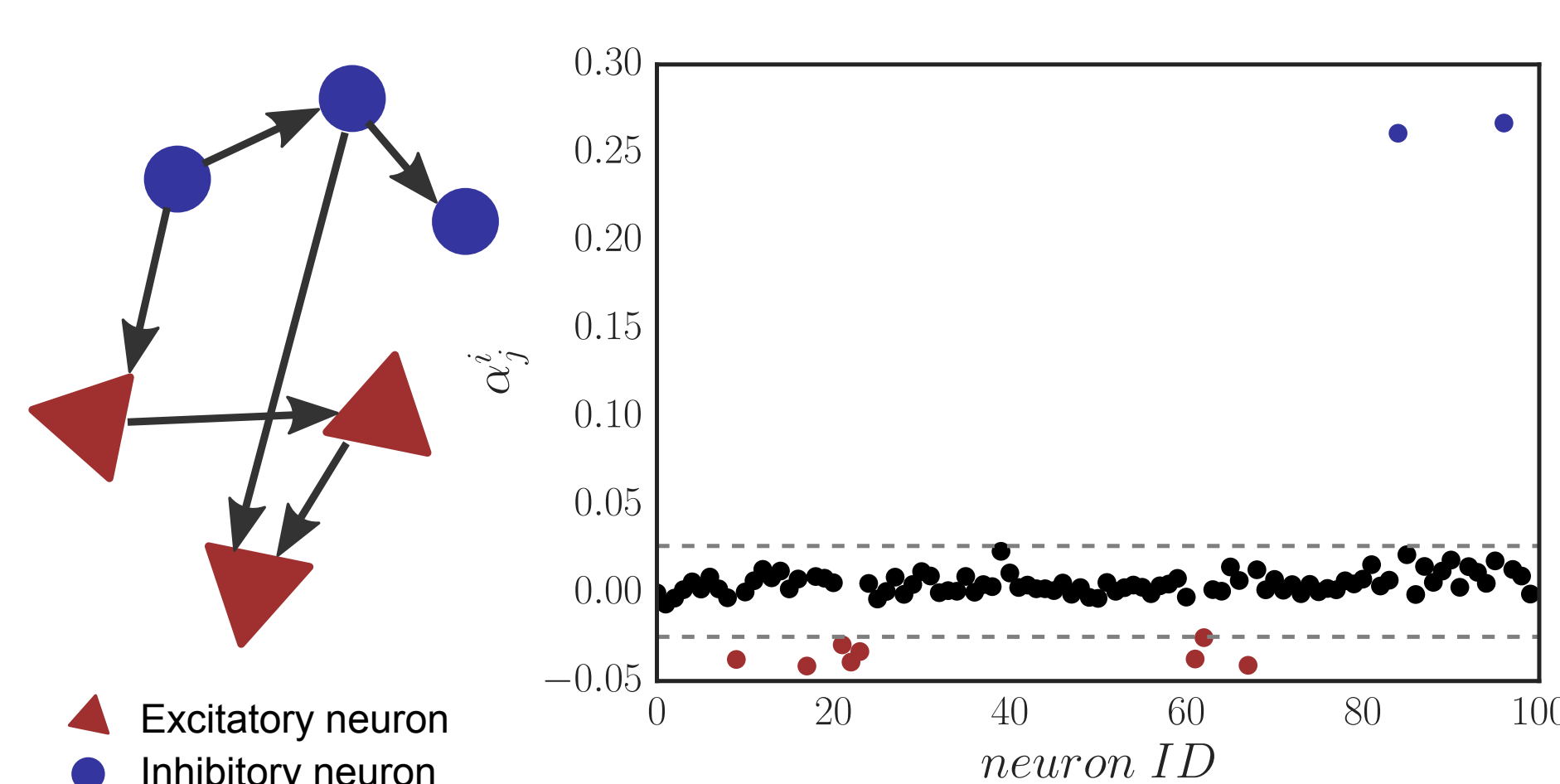
Dynamics of N-coupled neurons:

$$\Delta T^i = h_i(\Lambda^i w^i)$$

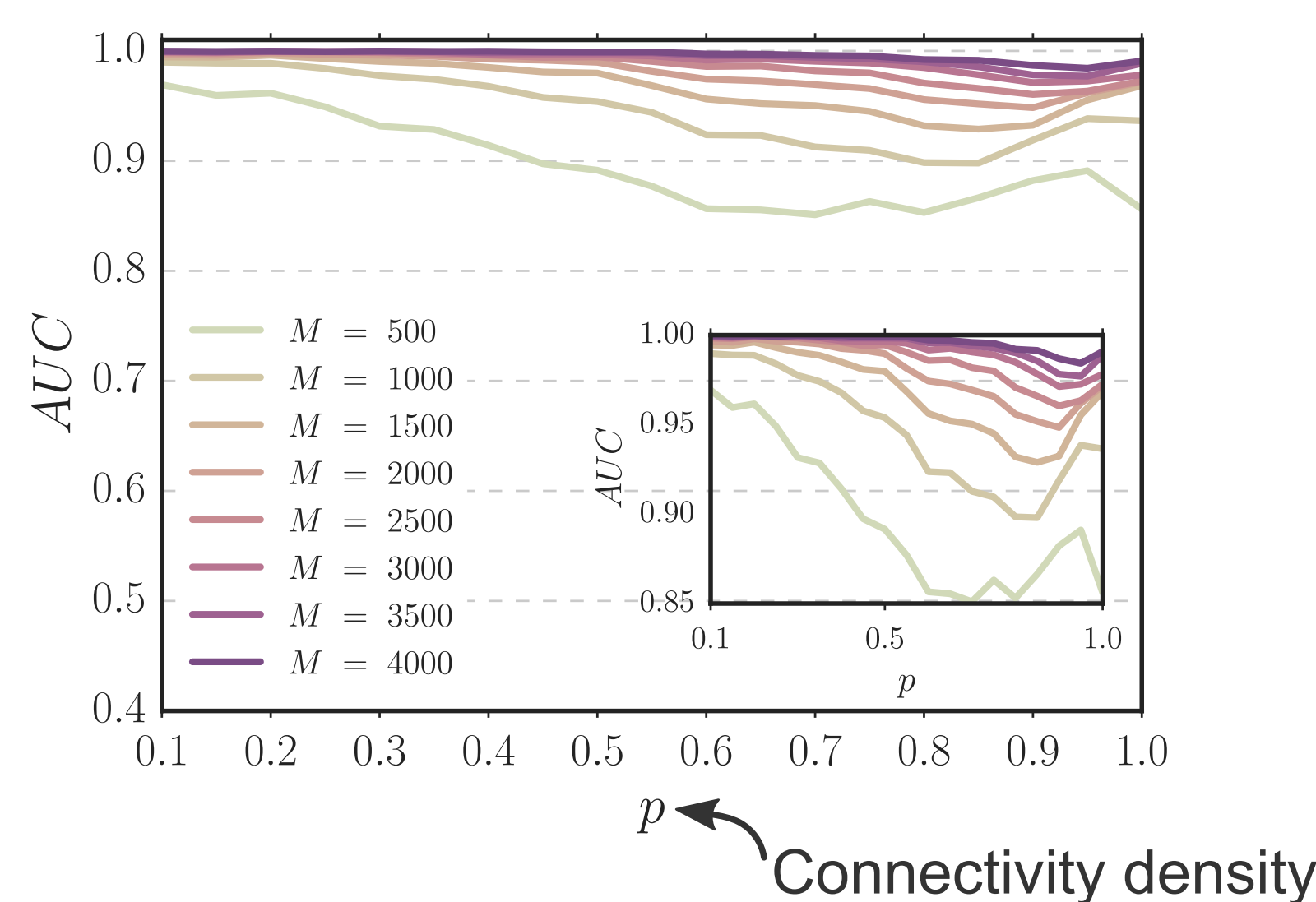
$$w^i = [w_1^i, w_2^i, \dots, w_N^i]^T \in \mathbb{R}^N$$

$$\Lambda_{jj}^i = \begin{cases} 0, & \text{if } \frac{\partial h_i}{\partial w_j} \equiv 0 \\ 1, & \text{if } \frac{\partial h_i}{\partial w_j} \neq 0 \end{cases}$$

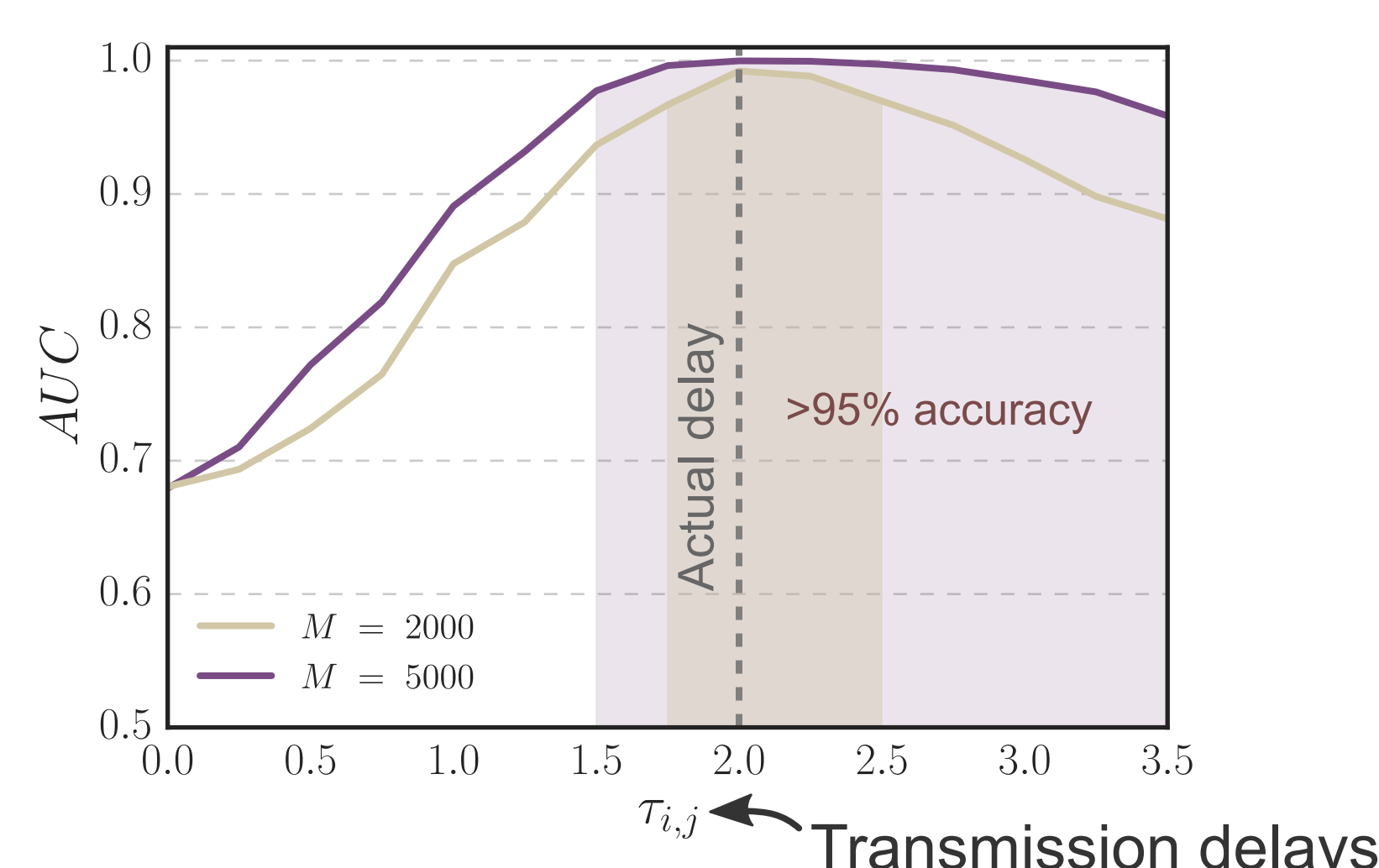
Inferring networks from spike trains:



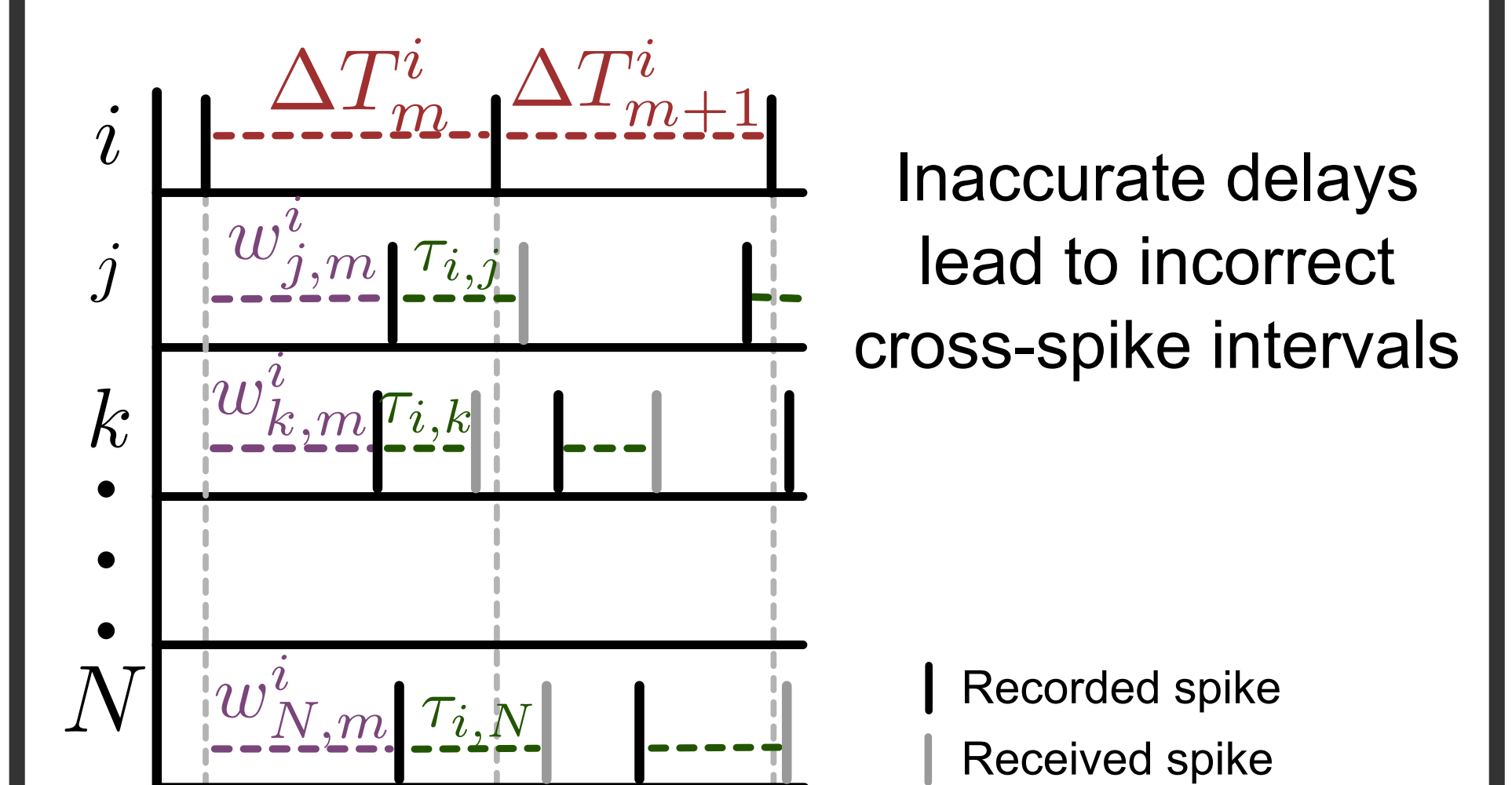
Accurate predictions also for dense networks



Correct predictions for erroneous delays



Inferring transmission delays:



Linear approximation error:

$$E(\tau_i) = \frac{\sum_m \|\Delta T_m^i - \nabla h_i \Lambda^i w_m^i\|_2}{\max_{\tau_i} \sum_m \|\Delta T_m^i - \nabla h_i \Lambda^i w_m^i\|_2}$$

Actual delay vector minimises the approximation error

