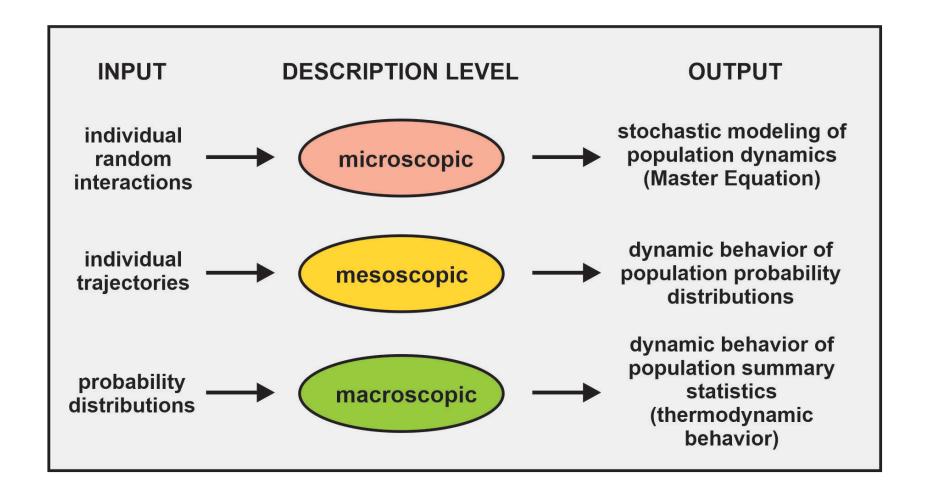
# LECTURE #13

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## Description Levels of Complex Systems



 To better understand what might happen at steady-state, assume that the population ME has a <u>unique</u> stationary solution

$$\overline{p}_{\mathbf{X}}(\mathbf{x};\Omega) \triangleq \lim_{t \to \infty} p_{\mathbf{X}}(\mathbf{x};t,\Omega)$$

that is <u>independent</u> of the initial state but depends on the size parameter  $\Omega$  in general.

The probability distribution  $p_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};t,\Omega)$  of the population density process  $\tilde{\mathbf{X}}(t;\Omega) = \mathbf{X}(t;\Omega)/\Omega$  is given by  $p_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};t,\Omega) = p_{\mathbf{X}}(\Omega\tilde{\mathbf{X}};t,\Omega)$ .

□ Let us define the <u>potential energy function</u>:

$$V(\tilde{\mathbf{x}}; \Omega) \triangleq -\frac{1}{\Omega} \ln \frac{\overline{p}_{\tilde{\mathbf{x}}}(\tilde{\mathbf{x}}; \Omega)}{\overline{p}_{\tilde{\mathbf{x}}}(\tilde{\mathbf{x}}_*; \Omega)} \geq 0$$

where

$$\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};\Omega) \triangleq \lim_{t \to \infty} p_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};t,\Omega)$$

and  $\tilde{\mathbf{X}}_*$  is a state at which the stationary probability distribution  $\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};\Omega)$  attains its maximum value.

Clearly,  $V(\tilde{\mathbf{x}};\Omega)$  assigns minimum (zero) potential to the states of maximum steady-state probability (which are referred to as **ground states**) and infinite potential to states of zero steady-state probability (*improbable states*).

$$V(\tilde{\mathbf{x}};\Omega) \triangleq -\frac{1}{\Omega} \ln \frac{\overline{p}_{\tilde{\mathbf{x}}}(\tilde{\mathbf{x}};\Omega)}{\overline{p}_{\tilde{\mathbf{x}}}(\tilde{\mathbf{x}}_*;\Omega)} \geq 0$$

The previous equation implies

$$\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};\Omega) = \frac{1}{\zeta(\Omega)} \exp\left\{-\Omega V(\tilde{\mathbf{X}};\Omega)\right\} \quad \text{GIBBS DISTRIBUTION}$$

where

$$\zeta(\Omega) \triangleq \sum_{\mathbf{u}} \exp\{-\Omega V(\mathbf{u};\Omega)\}$$
 PARTITION FUNCTION

- Assume that, close to the thermodynamic limit, the potential energy function  $V(\tilde{\mathbf{x}};\Omega)$  is an <u>analytic function</u> of  $\Omega^{-1}$ .
- lacktriangle Then, a Taylor series expansion with respect to  $\Omega^{-1}$  results in

$$V(\tilde{\mathbf{x}};\Omega) = V(\tilde{\mathbf{x}};\infty) + \frac{1}{\Omega} \frac{\partial V(\tilde{\mathbf{x}};\infty)}{\partial \Omega^{-1}} + \frac{1}{\Omega^2} \frac{\partial^2 V(\tilde{\mathbf{x}};\infty)}{\partial \Omega^{-2}} + \cdots$$
$$= V_0(\tilde{\mathbf{x}}) + \frac{1}{\Omega} V_1(\tilde{\mathbf{x}}) + \frac{1}{\Omega^2} V_2(\tilde{\mathbf{x}}) + \cdots$$

$$V(\tilde{\mathbf{x}};\Omega) = V_0(\tilde{\mathbf{x}}) + \frac{1}{\Omega}V_1(\tilde{\mathbf{x}}) + \frac{1}{\Omega^2}V_2(\tilde{\mathbf{x}}) + \cdots$$

$$V_{0}(\tilde{\mathbf{x}}) \triangleq V(\tilde{\mathbf{x}}; \infty) = -\lim_{\Omega \to \infty} \frac{1}{\Omega} \ln \frac{\overline{p}_{\tilde{\mathbf{x}}}(\tilde{\mathbf{x}})}{\overline{p}_{\tilde{\mathbf{x}}}(\tilde{\mathbf{x}}_{*})} \ge 0$$

$$V_1(\tilde{\mathbf{x}}) \triangleq \frac{\partial V(\tilde{\mathbf{x}}; \infty)}{\partial \Omega^{-1}}$$

$$V_2(\tilde{\mathbf{x}}) \triangleq \frac{\partial^2 V(\tilde{\mathbf{x}}; \infty)}{\partial \Omega^{-2}}$$

### Potential Energy and Stationary Distribution

□ In this case, we approximately have that

$$\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}}; \Omega) = \frac{1}{\zeta(\Omega)} \exp \left\{ -\Omega V_0(\tilde{\mathbf{X}}) - V_1(\tilde{\mathbf{X}}) - \Omega^{-1} V_2(\tilde{\mathbf{X}}) - \cdots \right\}$$

$$\zeta(\Omega) = \sum_{\mathbf{u}} \exp \left\{ -\Omega V_0(\mathbf{u}) - V_1(\mathbf{u}) - \Omega^{-1} V_2(\mathbf{u}) - \cdots \right\}$$

 $\Box$  If  $\chi(t)$  satisfies the macroscopic equations

$$\frac{d\chi_n(t)}{dt} = \sum_{m=1}^{M} s_{nm} \tilde{\pi}_m(\chi(t)), \quad t > 0, \ n = 1, 2, ..., N$$

then, it can be shown that

$$\frac{dV_0(\mathbf{\chi}(t))}{dt} = \sum_{n=1}^{N} \frac{\partial V_0(\mathbf{\chi}(t))}{\partial \mathbf{\chi}_n(t)} \frac{d\mathbf{\chi}_n(t)}{dt} \le 0$$

provided that  $V_0(\chi(t)) < \infty$ .

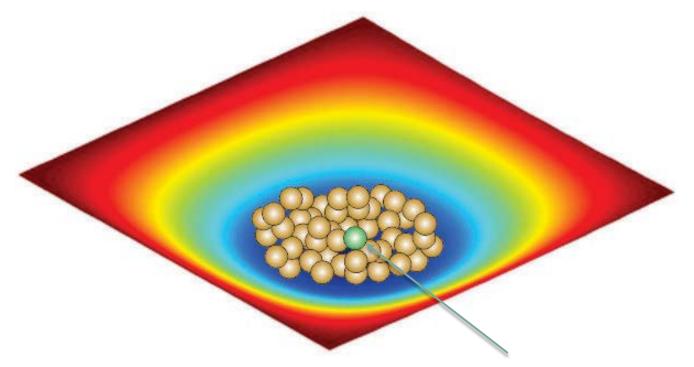
Consequently, the macroscopic solution produces dynamics that <u>never</u> increase the value of the potential energy function  $V_0$ .

$$\frac{dV_0(\mathbf{\chi}(t))}{dt} = \sum_{n=1}^{N} \frac{\partial V_0(\mathbf{\chi}(t))}{\partial \mathbf{\chi}_n(t)} \frac{d\mathbf{\chi}_n(t)}{dt} < 0$$

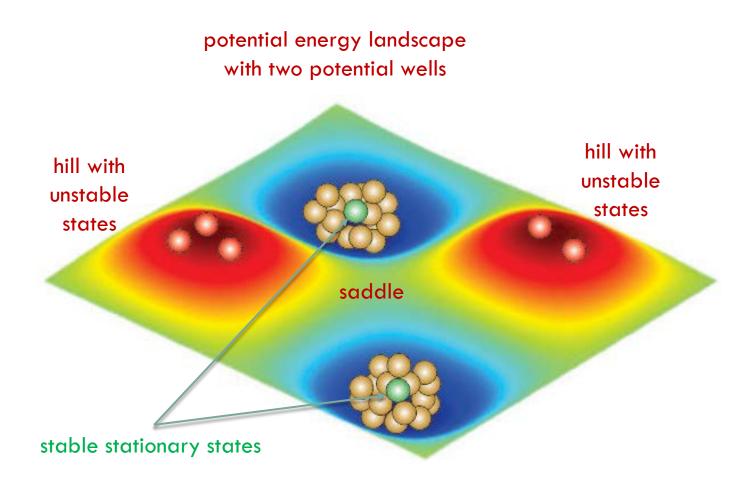
The previous result implies that, if  $\chi'$  is a <u>local minimum</u> of  $V_0$ , then the macroscopic solution will always <u>converge</u> to  $\chi'$ , provided that the macroscopic system is initialized by a state that is also near  $\chi'$  (this implies that  $\chi'$  is an asymptotically <u>stable solution</u> of the macroscopic equations).

- We can view the multidimensional surface  $V_0(\tilde{\mathbf{x}})$  as a <u>potential</u> energy landscape.
- The <u>stable stationary states</u> of the macroscopic equations correspond to <u>potential wells</u> (basins of attraction) associated with the minima of  $V_0(\tilde{\mathbf{x}})$  separated by barriers, corresponding to <u>hills</u> associated with <u>unstable states</u>, and <u>saddles</u> associated with <u>transitional states</u> (states on the potential energy surface from which stable states are equally accessible).
- Which path the macroscopic system takes along the potential energy landscape will depend on the initial condition.

potential energy landscape with one potential well



stable stationary state



- Initialization within a potential well guarantees that the macroscopic dynamics will stay within the well <u>permanently</u>.
- If the macroscopic system reaches a minimum of the potential energy landscape, then this minimum must be a stationary state of the macroscopic system since uphill motions are <u>not possible</u> since

$$\frac{dV_0(\mathbf{\chi}(t))}{dt} \le 0$$

If the macroscopic system reaches a minimum of the potential energy landscape, it stays there forever!!

□ For large  $\Omega$ , we approximately have  $V(\tilde{\mathbf{x}};\Omega) = V_0(\tilde{\mathbf{x}}) + \Omega^{-1}V_1(\tilde{\mathbf{x}})$ , in which case

$$\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}}; \Omega) = \frac{1}{\zeta(\Omega)} \exp\left\{-\Omega V_0(\tilde{\mathbf{X}}) - V_1(\tilde{\mathbf{X}})\right\}$$
$$\zeta(\Omega) = \sum_{\mathbf{u}} \exp\left\{-\Omega V_0(\mathbf{u}) - V_1(\mathbf{u})\right\}$$

□ In this case, we can show that

$$\lim_{\Omega \to \infty} \overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}}; \Omega) = \begin{cases} \frac{\exp\left\{-V_1(\tilde{\mathbf{x}})\right\}}{\sum_{\mathbf{u} \in \mathbb{S}_0} \exp\left\{-V_1(\mathbf{u})\right\}}, & \text{for } \tilde{\mathbf{x}} \in \mathbb{S}_0 \\ 0, & \text{for } \tilde{\mathbf{x}} \notin \mathbb{S}_0 \end{cases} \quad \text{ground states of } V_0$$

see supplement #8 for details

$$\lim_{\Omega \to \infty} \overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}};\Omega) = \begin{cases} \frac{\exp\left\{-V_1(\tilde{\mathbf{x}})\right\}}{\sum_{\mathbf{u} \in \mathbb{S}_0} \exp\left\{-V_1(\mathbf{u})\right\}}, & \text{for } \tilde{\mathbf{x}} \in \mathbb{S}_0 \\ 0, & \text{for } \tilde{\mathbf{x}} \notin \mathbb{S}_0 \end{cases} \quad \text{ground}$$

- In the <u>thermodynamic limit</u>, the ME asymptotically converges almost surely (with probability one) to a ground state of the potential energy function  $V_0$ , <u>independently</u> of the initial state.
- The specific ground state is chosen with probability determined by the values of the potential energy function  $V_1$  over the ground states of  $V_0$ .
- On the other hand, the macroscopic equations might reach a minimum of  $V_0$  which may or may not be a ground state, depending on the initial condition.

### Keizer's Paradox

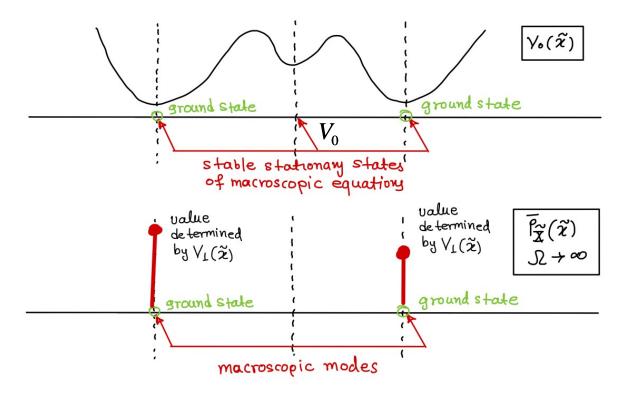
- If the macroscopic equations have a <u>unique</u> stable solution at steady-state that is independent of the initial condition, then  $V_0$  will have <u>only one</u> (global) minimum.
- In this case, the ME will converge almost surely (with probability one) to the same state in the thermodynamic limit.
- However, if  $V_0$  contains more than one (global or local) minimum, then the stationary solution of the ME may be different from the stationary solution predicted by the corresponding macroscopic equations.
- Consequently,

$$\lim_{\Omega \to \infty} \lim_{t \to \infty} p_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}}; t, \Omega) \neq \lim_{t \to \infty} \lim_{\Omega \to \infty} p_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}}; t, \Omega)$$

This distinct difference between the stationary behavior of the ME (left-hand side of inequality) and of the macroscopic equations (right-hand side of inequality) is known as <u>Keizer's paradox</u>.

- In the thermodynamic limit as  $\Omega \to \infty$ , the peaks present in the stationary probability distribution  $\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}};\infty)$  will be associated only with the <u>global</u> minima of  $V_0$ , which are in turn associated with stable stationary states of the macroscopic equations.
- extstyle ext
- Note however that there might be stable stationary states of the macroscopic equations that <u>do not</u> introduce peaks in the stationary probability distribution  $\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}};\infty)$  and will therefore be deemed to be improbable according to the master equation approach.
- $\square$  These states are associated with the <u>local</u> minima of  $V_0$ .

### **Example:**

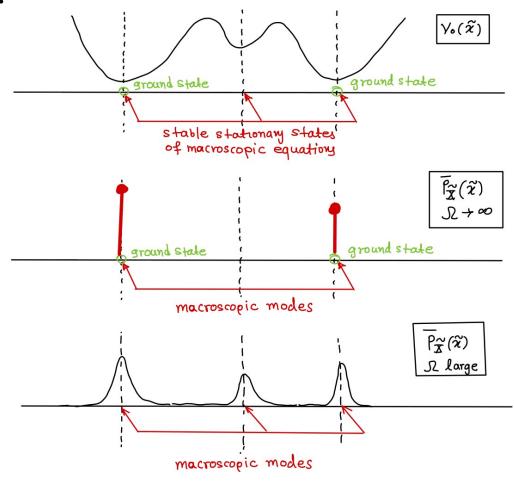


$$\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}};\Omega) = \frac{1}{\zeta(\Omega)} \exp\left\{-\Omega V_0(\tilde{\mathbf{x}}) - V_1(\tilde{\mathbf{x}})\right\}$$

- At finite but sufficiently large sizes  $\Omega$ , the <u>peaks</u> of the stationary probability distribution  $\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};\Omega)$  will correspond to <u>minima</u> of the potential energy landscape  $V_0(\tilde{\mathbf{X}}) + \Omega^{-1}V_1(\tilde{\mathbf{X}})$ .
- In addition, and for large enough  $\Omega$ , we have that  $V_0(\tilde{\mathbf{x}}) + \Omega^{-1}V_1(\tilde{\mathbf{x}}) \simeq V_0(\tilde{\mathbf{x}})$ .
- We therefore expect in this case that  $\tilde{\mathbf{x}}_0$  will be a minimum of  $V_0(\tilde{\mathbf{x}}) + \Omega^{-1}V_1(\tilde{\mathbf{x}})$  if and only if it is a minimum of  $V_0(\tilde{\mathbf{x}})$  and the peaks of the stationary probability distribution  $\overline{p}_{\tilde{\mathbf{x}}}(\tilde{\mathbf{x}};\Omega)$  will correspond to the <u>stable</u> stationary states of the macroscopic equations.

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peaks in \overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}};\Omega) \Leftrightarrow \min \text{ in } V_0(\tilde{\mathbf{x}}) + \Omega^{-1}V_1(\tilde{\mathbf{x}}) \text{ [for sufficiently large but finite } \Omega]
\Leftrightarrow \min \text{ in } V_0(\tilde{\mathbf{x}}) \text{ [for sufficiently large but finite } \Omega]
\Leftrightarrow \text{ stable macroscopic states at steady-state}
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### Example:



 $\hfill\Box$  At smaller values of  $\Omega$  , the stationary probability distribution will be given by

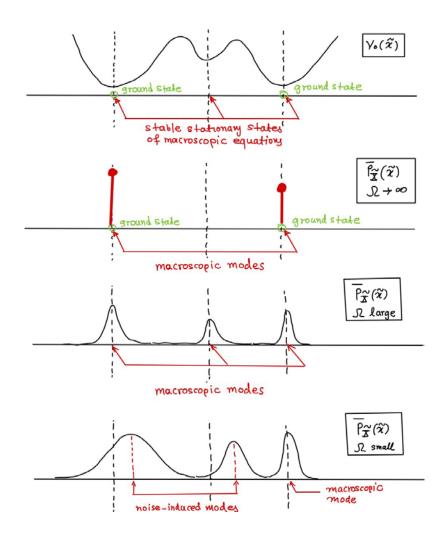
$$\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}};\Omega) = \frac{1}{\zeta(\Omega)} \exp\left\{-\Omega V_0(\tilde{\mathbf{x}}) - V_1(\tilde{\mathbf{x}}) - \Omega^{-1}V_2(\tilde{\mathbf{x}}) - \cdots\right\}$$

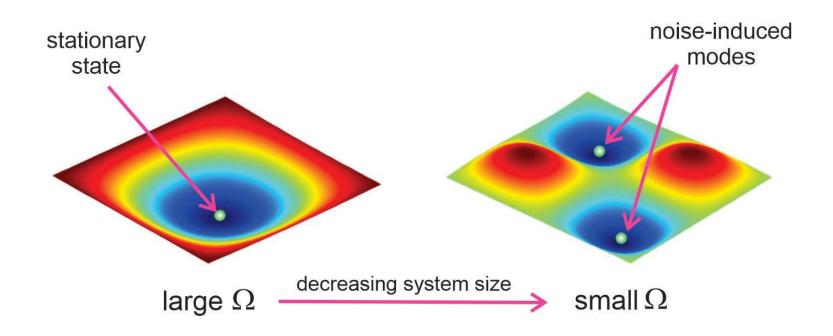
- The modes of  $\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};\Omega)$  will now be determined by the minima of the potential energy landscape  $V(\tilde{\mathbf{X}};\Omega) = V_0(\tilde{\mathbf{X}}) + \Omega^{-1}V_1(\tilde{\mathbf{X}}) + \Omega^{-2}V_2(\tilde{\mathbf{X}}) + \cdots$ .
- However, a state that minimizes the potential energy function V may not necessarily minimize  $V_0$ , in which case at least some modes of the probability distribution  $\overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{x}};\Omega)$  will not be predicted by the corresponding macroscopic equations.

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peaks in \overline{p}_{\tilde{\mathbf{X}}}(\tilde{\mathbf{X}};\Omega) \Leftrightarrow \min \text{ in } V(\tilde{\mathbf{X}};\Omega) = V_0(\tilde{\mathbf{X}}) + \Omega^{-1}V_1(\tilde{\mathbf{X}}) + \Omega^{-2}V_2(\tilde{\mathbf{X}}) + \cdots
\Leftrightarrow \min \text{ of } V_0(\tilde{\mathbf{X}})
\Leftrightarrow \text{ stable macroscopic states at steady-state}
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- These modes are referred to as noise-induced modes, since they show up at small system sizes in which appreciable stochastic fluctuations may be present in the system due to stochasticity intrinsic to the system.
- The presence of noise-induced modes in nonlinear reaction networks and their importance in modeling system behavior not accounted for by their macroscopic counterparts has been welldocumented in the literature (especially in biology).

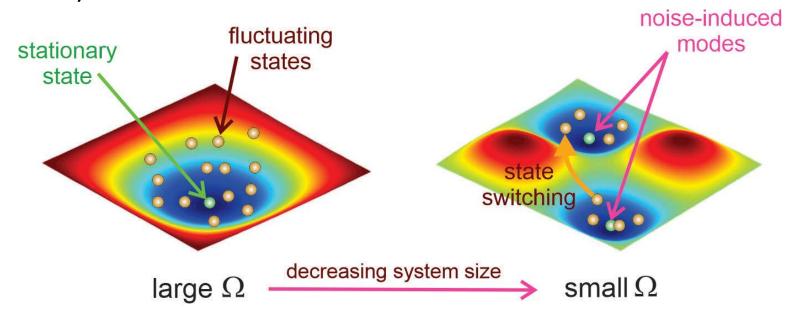
### **Example:**





### Mode Switching

- If a Markovian reaction network is at a stable state, intrinsic stochasticity may force it to switch to another stable state at a later time.
- The probability of switching from a stable state to another stable state tends (in general exponentially) to zero as the system size increases to infinity.



### Mode Switching

- $\hfill\Box$  At finite system sizes  $\Omega$  , switching among stable stationary states becomes possible, but the probability of switching is very small for large  $\Omega$  .
- Switching among stable stationary states is usually a <u>rare event (it</u> happens with extremely small probability).
- As a matter of fact, the waiting time for switching can be approximated by an exponential distribution with a rate parameter that tends to zero in the thermodynamic limit as  $\Omega \to \infty$ .
- Efficient switching between modes requires small system sizes and thus appreciable noise.