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Quasi-stationary measures and metastability

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Outline

1. Introduction

- Metastable systems.
- Markovian models.
- Metastable state: restricted ensemble and quasi stationary measure

2. Exit time: law and sharp average estimates

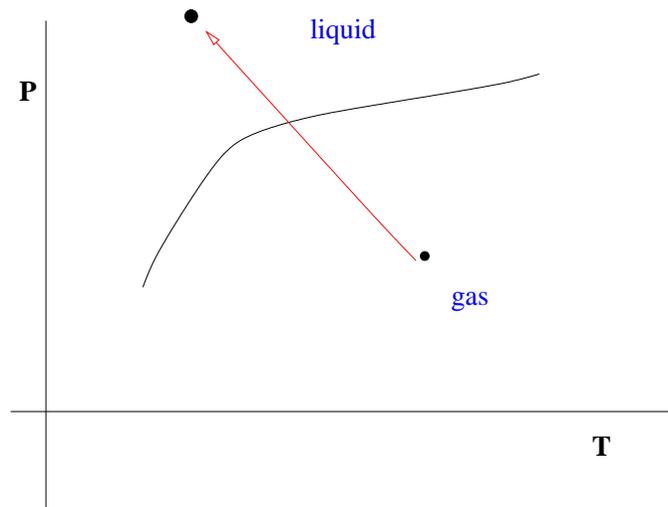
- Exponential law of the exit time.
- Sharp estimates on average exit time and relaxation time.
- Example: Curie-Weiss model.

3. Escape from metastability

- Soft measures as generalization of quasi-stationary measures.
- Transition times and mixing time asymptotics.

Metastable systems

Metastability is a common dynamical phenomenon related to *first order phase transition*.



If the parameters of the system change along the line of the first order phase transition, *the system moves from one metastable state to the new equilibrium*.

Main features: This transition takes a **long time**, while the system stays in an **apparent equilibrium**.

Rigorous description

Due to the work of Lebowitz & Penrose (*J. Stat. Phys.*, **3**, 1971):

”We shall characterize metastable thermodynamic states by the following properties:

- (a) only one thermodynamic phase is present,*
- (b) a system that starts in this state is likely to take a long time to get out,*
- (c) once the system has gotten out, it is unlikely to return. ”*

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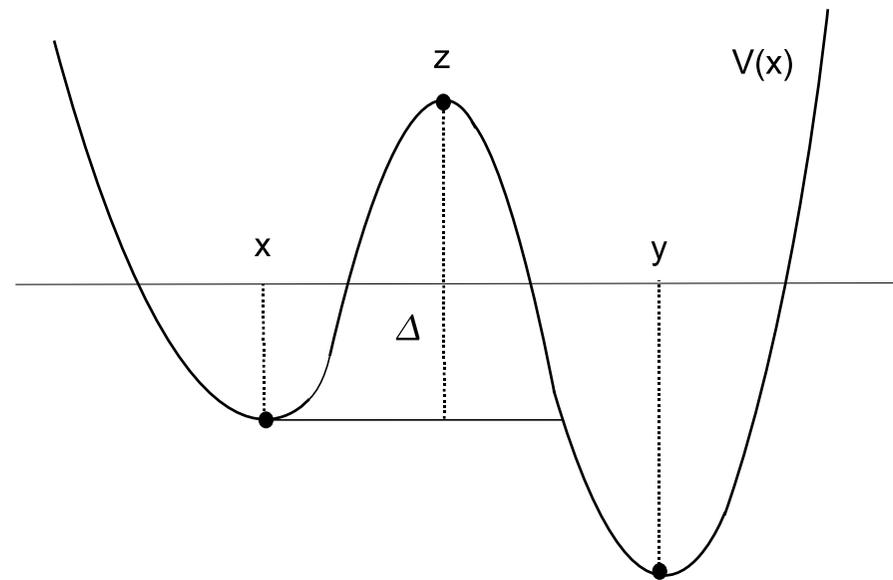
one phase of metastable state \longrightarrow region $\mathcal{R} \subset \mathcal{X}$ of the phase space
metastable state $\longrightarrow \mu_{\mathcal{R}} = \mu(\cdot | \mathcal{R})$, the restricted ensemble.

Main question: Show properties (b) and (c) by analyzing the exit time from \mathcal{R} : $\mathcal{T}_{\mathcal{R}^c}$.

Metastability in stochastic dynamics

Previous results and techniques

A simple example: Let $X_t \in \mathbb{R}$ solution of $dX_t = -V'(X_t) + \sqrt{2\varepsilon} dW_t$



Metastable systems

- **Large deviations techniques** [Freidlin, Wentzell ('84)]:

$$(1) \quad \lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbb{E}_x \mathcal{T}_y = \Delta \quad (2) \quad \lim_{\varepsilon \rightarrow 0} \mathbb{P}_x \left(\frac{\mathcal{T}_y}{\mathbb{E}_x \mathcal{T}_y} > t \right) = e^{-t}$$

- **Pathwise approach** [Cassandro, Galves, Olivieri, Vares ('84)]:

It focuses on typical trajectories and exponential law of the exit time.

By LD techniques, it provides (1)-(2). Developed and generalized in many ways: [Neves, Schonmann ('92)], [Ben Arous, Cerf ('96)], [Schonmann, Shlosman ('98)], [Gaudillière, Olivieri, Scoppola ('05)].

- **Potential theoretic approach** [Bovier, Eckhoff, Gaynard, Klein ('01-'04)]:

It focuses on relation between exit time and capacities, (and spectrum of the generator), providing sharp results (T finite): [Bovier, Manzo ('02)] [B., Bovier, Ioffe, '09], [Bovier, Den Hollander, Spitoni ('10)], [Beltrán, Landim ('10)].

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Our main goal: *Give a different description of metastable state and find simple hypotheses to get sharp estimates on the average exit time and prove its exponential law.*

Markovian Models

Markov process $X = (X_t)_{t \in \mathbb{R}}$ on a finite set \mathcal{X} with generator

$$\mathcal{L}f(x) = \sum_{y \in \mathcal{X}} p(x, y)(f(y) - f(x))$$

For $\mathcal{R} \subset \mathcal{X}$ **metastable set**, let $X_{\mathcal{R}}$ ($X_{\mathcal{R}^c}$) be the reflected process on \mathcal{R} (\mathcal{R}^c).

Assume:

- 1) X *irreducible* and *reversible w.r.t. μ* ;
- 2) $X_{\mathcal{R}}$, $X_{\mathcal{R}^c}$ *irreducible* \longrightarrow *reversible w.r.t. $\mu_{\mathcal{R}}$ and $\mu_{\mathcal{R}^c}$* .

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- Consider the **sub-Markovian kernel** on \mathcal{R}

$$r^*(x, y) = p(x, y), \quad \text{for all } x, y \in \mathcal{R}$$

and let

$$e_{\mathcal{R}}(x) = \sum_{y \notin \mathcal{R}} p(x, y) \quad (\text{escape probability from } \mathcal{R}).$$

Quasi-stationary measure

From Perron-Frobenius Theorem and Darroch & Seneta('62):

- \exists a **measure** $\mu_{\mathcal{R}}^*$ on \mathcal{R} , called **quasi stationary measure** defined as

$$\mu_{\mathcal{R}}^*(y) = \lim_{t \rightarrow \infty} \mathbb{P}_x(X(t) = y | \mathcal{T}_{\mathcal{R}^c} > t) \quad \text{Yaglom limit}$$

- Moreover $\exists \phi^* > 0$ s.t.

1. $\mu_{\mathcal{R}}^* r^* = (1 - \phi^*) \mu_{\mathcal{R}}^* \longrightarrow$ **left eigenvector**

2. $\mathbb{P}_{\mu_{\mathcal{R}}^*}(\mathcal{T}_{\mathcal{R}^c} > t) = e^{-\phi^* t} \longrightarrow$ **exponential law**

3. $\mathbb{E}_{\mu_{\mathcal{R}}^*}(\mathcal{T}_{\mathcal{R}^c})^{-1} = \phi^* = \mu_{\mathcal{R}}^*(e_{\mathcal{R}}) \longrightarrow$ **exponential rate** .

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- Choose $\mu_{\mathcal{R}}^*$ instead of $\mu_{\mathcal{R}}$ in order to describe the metastable state.

Quasi-stationary measure

Advantages and disadvantages.

- $\mu_{\mathcal{R}}^*$ immediately provides the exponential law of $\mathcal{T}_{\mathcal{R}}$, that in general is hard to deduce.
- $\mu_{\mathcal{R}}^*$ is not explicitly given, then preventing from getting quantitative estimates.

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Let $\gamma_{\mathcal{R}}$ be the **spectral gap** of $X_{\mathcal{R}}$ and define $\varepsilon_{\mathcal{R}} := \frac{\phi^*}{\gamma_{\mathcal{R}}}$.

Proposition 1. *If $\varepsilon_{\mathcal{R}} < 1$, then*

$$\left\| \frac{\mu_{\mathcal{R}}^*}{\mu_{\mathcal{R}}} - 1 \right\|_{\mathcal{R},2}^2 \leq \frac{\varepsilon_{\mathcal{R}}}{1 - \varepsilon_{\mathcal{R}}}$$

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Remark. *Note that $\varepsilon_{\mathcal{R}} = \gamma_{\mathcal{R}}^{-1} / \mathbb{E}_{\mu_{\mathcal{R}}^*}(\mathcal{T}_{\mathcal{R}^c})$.*

For metastable systems, we expect $\varepsilon_{\mathcal{R}} \ll 1$ with some parameter of the system (e.g. size of the system $\rightarrow \infty$, $T \rightarrow 0$)

Exponential law of the exit time

Assume that $\varepsilon_{\mathcal{R}} \rightarrow 0$ and let $S_{\mathcal{R}} := \frac{1}{\gamma_{\mathcal{R}}^*} \ln \frac{2}{\delta(1-\delta)\zeta_{\mathcal{R}}}$ (local mixing time),
with $\zeta_{\mathcal{R}} := \min_{x \in \mathcal{R}} \{\mu_{\mathcal{R}}^*(x) / \mu_{\mathcal{R}}(x)\}$, $\gamma_{\mathcal{R}}^*$ the spectral gap of r^* , and $\delta = O(\varepsilon_{\mathcal{R}})$.

THM 1. [Exponential law] *If $S_{\mathcal{R}} \cdot \phi^* = o(1)$ as $\varepsilon_{\mathcal{R}} \rightarrow 0$, then*

$$1) \quad \mathbb{E}_{\mu_{\mathcal{R}}}(\mathcal{T}_{\mathcal{R}^c}) = \phi^{*-1}(1 + o(1))$$

$$2) \quad \mathbb{P}_{\mu_{\mathcal{R}}}(\mathcal{T}_{\mathcal{R}^c} > t \cdot \phi^{*-1}) = e^{-t}(1 + o(1))$$

Remark. *In fact we can prove much more. We can consider general initial measure ν , and get exact corrective terms which are matching in the regime $S_{\mathcal{R}} \cdot \phi^* = o(1)$.*

Sharp average estimates

Recall that:

$$\text{If } A, B \subset \mathcal{X}, A \cap B = \emptyset \implies \text{cap}(A, B) = \sum_{a \in A} \mu(a) \mathbb{P}_a(\tau_A^+ > \tau_B^+).$$

As shown in a series of papers by Bovier, Eckhoff, Gayrard & Klein ('01-'04), *capacities enter in the computation of the average exit time from A to B.*

Main advantage of capacities, is that **they satisfy a two-sided variational principle**

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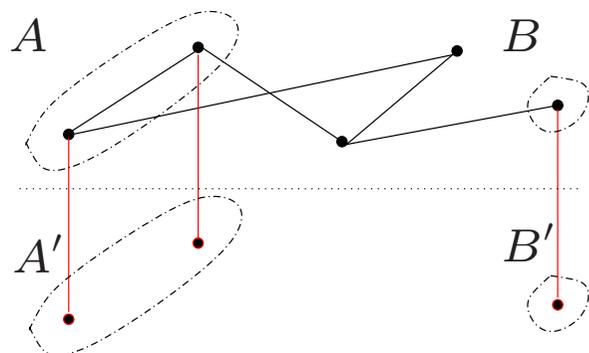
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Generalized capacities

For $k, \lambda > 0$, define an **extended system** $\mathcal{X}' = \mathcal{X} \cup A' \cup B'$, A', B' copies of A, B .



$$c'(x, y) = c(x, y) = \mu(x)p(x, y)$$

$$c'(a, a') = k\mu(a)$$

$$c'(b, b') = \lambda\mu(b)$$

Exit time: law and sharp average estimates

Definition (k, λ -capacities): $\text{cap}_k^\lambda(A, B) = \text{cap}(A', B')$.

When $\lambda = +\infty \longrightarrow B = B'$ and $\text{cap}_k^\infty(A, B) = \text{cap}_k(A, B)$.

In particular $\text{cap}_\infty^\infty(A, B) = \text{cap}(A, B)$.

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THM 2. [Mean exit time] *If $S_{\mathcal{R}} \cdot \phi^* = o(1)$ as $\varepsilon_{\mathcal{R}} \rightarrow 0$, and choosing $\phi^* \ll k \ll \gamma_{\mathcal{R}}$,*

$$\phi^{*-1} = \frac{\mu(\mathcal{R})}{\text{cap}_k(\mathcal{R}, \mathcal{R}^c)}(1 + o(1))$$

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THM 3. [relaxation time] *If $S_{\mathcal{R}} \cdot \phi^* = o(1)$ and $S_{\mathcal{R}^c} \cdot \phi^{c*} = o(1)$ with $\varepsilon_{\mathcal{R}}, \varepsilon_{\mathcal{R}^c} \rightarrow 0$, and choosing $\phi^* \ll k \ll \gamma_{\mathcal{R}}$ and $\phi^{c*} \ll \lambda \ll \gamma_{\mathcal{R}^c}$, then*

$$\mathcal{T}_{rel} \equiv \frac{1}{\gamma} = \frac{\mu(\mathcal{R})\mu(\mathcal{R}^c)}{\text{cap}_k^\lambda(\mathcal{R}, \mathcal{R}^c)}(1 + o(1))$$

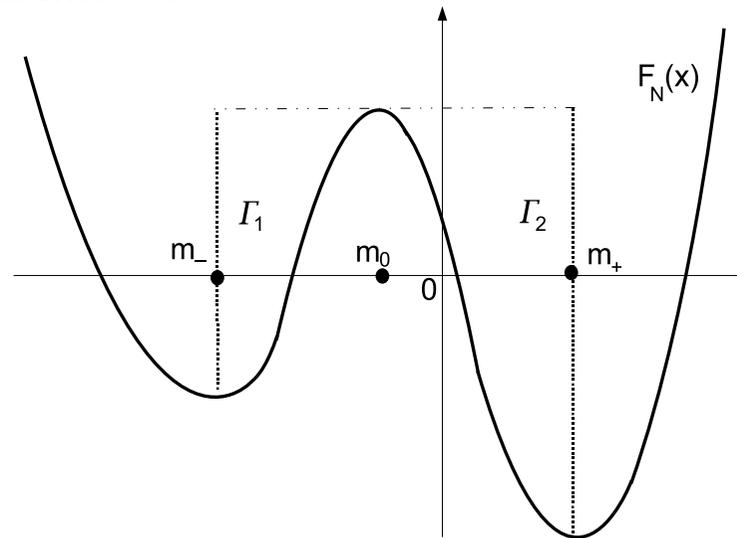
Example: Curie-Weiss model

A simple example: the Curie-Weiss model

Let $m \in \Gamma = \{-1, -1 + \frac{2}{N}, \dots, 1\}$ (magnetization) a 1D-parameter

Let $\mu(m) \propto e^{-\beta N F_N(m)}$ the Gibbs measure on Γ and consider a dynamics reversible w.r.t. μ with transition rates $p(m, m^\pm) \propto e^{-\beta N \nabla_\pm F_N}$.

For some values of the parameters



Let $\mathcal{R} = \{\sigma \in \mathcal{X} : m_N(\sigma) \leq m_0\}$.

Example: Curie-Weiss model

Questions:

1. Law and average of $\mathcal{T}_{\mathcal{R}^c}$ w.r.t. $\mu_{\mathcal{R}}$?
2. Relaxation time?

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First step: verify the hypotheses

We want to show that $\varepsilon_{\mathcal{R}}, \varepsilon_{\mathcal{R}^c} \xrightarrow{N \rightarrow \infty} 0$ and $S_{\mathcal{R}} \cdot \phi^* = o(1), S_{\mathcal{R}^c} \cdot \phi^{c^*} = o(1)$.

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and similarly $\phi^{c*} \leq e^{-\beta N \Gamma_2}$, with $\Gamma_1 < \Gamma_2$.

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2. $\gamma_{\mathcal{R}}^{-1} \leq \mathcal{T}_{mix}^{\mathcal{R}} \leq c(\beta)N^{3/2}$ ← argument used in [Levin, Luczak, Peres ('10)] .

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3. With the above estimates we get easily $S_{\mathcal{R}}, S_{\mathcal{R}^c} \leq c(\beta) N^3$.

→ Then the required hypotheses follow.

Example: Curie-Weiss model

Second step: compute the capacities

We make use of the **two-side variational principle** over the capacities.

Test functions and flows are provided by the **1D process over the magnetizations**, where capacities can be computed explicitly.

Then, for all $\phi_{\mathcal{R}}^* \ll k \ll \gamma_{\mathcal{R}}$ and $\phi_{\mathcal{R}^c}^* \ll \lambda \ll \gamma_{\mathcal{R}^c}$

$$1. \text{ cap}_k(\mathcal{R}, \mathcal{R}^c) = \frac{1}{Z_N} \cdot \frac{1}{\sqrt{\pi N}} c(m_0) e^{-\beta N f_N(m_0)} (1 + o(1)),$$

$$2. \text{ cap}_k^\lambda(\mathcal{R}, \mathcal{R}^c) = \frac{1}{Z_N} \cdot \frac{1}{2\sqrt{\pi N}} c(m_0) e^{-\beta N f_N(m_0)} (1 + o(1)),$$

where $c(m_0) = \sqrt{(1 - m_0^2) |f_N''(m_0)|}$.

Example: Curie-Weiss model

The result

From Theorems 1.,2. and 3., it holds

(i) $\mathcal{T}_{\mathcal{R}^c}$ has asymptotic **exponential law** w.r.t. $\mu_{\mathcal{R}}$ with **mean**

$$\mathbb{E}_{\mu_{\mathcal{R}}}(\mathcal{T}_{\mathcal{R}^c}) = \frac{\pi N}{\beta c(m_0)c(m_-)} e^{\beta N \Gamma_1} (1 + o(1))$$

(ii) The **relaxation time** γ^{-1} is given by

$$\gamma^{-1} = \frac{2\pi N}{\beta c(m_0)c(m_-)} e^{\beta N \Gamma_1} (1 + o(1))$$

Soft measure and escape from metastability

Recall property (c) of Lebowitz & Penrose:

”once the system has gotten out, it is unlikely to return ”

What does it mean ”to get out” from \mathcal{R} ? Exit from \mathcal{R} ?

When the system just exited \mathcal{R} , the probabilities to go back to \mathcal{R} or proceed in \mathcal{R}^c are equal, and (c) fails.

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Main Idea

If the dynamics spends in \mathcal{R}^c a time $\geq S_{\mathcal{R}^c}$ (local mixing in \mathcal{R}^c) then it is close to $\mu_{\mathcal{R}^c}^*$.

⇒

Define the ”true escape from \mathcal{R} ” as the first time that the ”dynamics on \mathcal{R} ” makes an excursion in \mathcal{R}^c of order $\geq S_{\mathcal{R}^c}$.

Soft measure and escape from metastability

Formally:

- For any $\lambda > 0$ and $\sigma_\lambda \sim \exp(\lambda)$ indep. of X , sub-Markovian kernel on \mathcal{R} :

$$r_\lambda^*(x, y) = \mathbb{P}_x(\mathbf{X}(\tau_{\mathcal{R}}^+) = y, L_{\mathcal{R}^c}(\tau_{\mathcal{R}}^+) \leq \sigma_\lambda)$$

where L_A = local time in $A \subset \mathcal{X}$ and G_A its right-continuous inverse.

Soft measure and escape from metastability

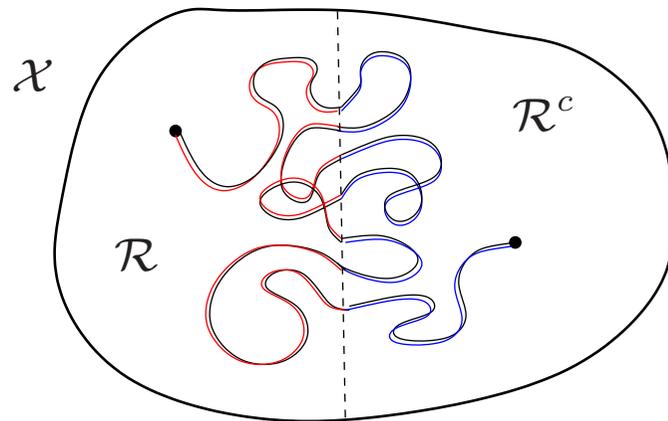
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where L_A = local time in $A \subset \mathcal{X}$ and G_A its right-continuous inverse.

- Define the **transition time**: $\mathcal{T}_{\mathcal{R}^c, \lambda} = L_{\mathcal{R}}(G_{\mathcal{R}^c}(\sigma_\lambda))$



σ_λ = length of blue-path

$G_{\mathcal{R}^c}(\sigma_\lambda)$ = length of black-path

$\mathcal{T}_{\mathcal{R}^c, \lambda}$ = length of red-path

Soft measure and escape from metastability

By similar arguments to those used for the analysis of r^* , we define the **soft measure** $\mu_{\mathcal{R},\lambda}^*$ on \mathcal{R} as

$$\mu_{\mathcal{R},\lambda}^*(y) = \lim_{t \rightarrow \infty} \mathbb{P}_x(X(G_{\mathcal{R}}(t)) = y | \mathcal{T}_{\mathcal{R}^c,\lambda} > t)$$

It turns out that $\exists \phi_{\lambda}^* > 0$ s.t.

1. $\mu_{\mathcal{R},\lambda}^* r_{\lambda}^* = (1 - \phi_{\lambda}^*) \mu_{\mathcal{R},\lambda}^* \longrightarrow$ *left eigenvector*
2. $\mathbb{P}_{\mu_{\mathcal{R},\lambda}^*}(\mathcal{T}_{\mathcal{R}^c,\lambda} > t) = e^{-\phi_{\lambda}^* t} \longrightarrow$ *exponential law*
3. $\mathbb{E}_{\mu_{\mathcal{R},\lambda}^*}(\mathcal{T}_{\mathcal{R}^c,\lambda})^{-1} = \phi_{\lambda}^* = \mu_{\mathcal{R},\lambda}^*(e_{\mathcal{R},\lambda}) \longrightarrow$ *average time*

Remark 1. $\mu_{\mathcal{R},\lambda}^*$ is *continuous interpolation* between $\mu_{\mathcal{R}} = \mu_{\mathcal{R},0}^*$ and $\mu_{\mathcal{R}}^* = \mu_{\mathcal{R},\infty}^*$.

Remark 2. The same construction can be done for the dynamics on \mathcal{R}^c : For $k > 0$ and taking a time (\mathcal{R}) -excursion bound of $\sigma_k \sim \exp(k)$, we construct $\mu_{\mathcal{R}^c,k}^*$.

Transition time and mixing time

THM 4. *All the results proved for $\mathcal{T}_{\mathcal{R}^c}$ and ϕ^* , hold for $\mathcal{T}_{\mathcal{R}^c, \lambda}$ and ϕ_λ^* under analogous hypotheses ($\varepsilon_{\mathcal{R}} \ll 1$ and $S_{\mathcal{R}, \lambda} \cdot \phi_\lambda^* = o(1)$ as $\varepsilon_{\mathcal{R}} \rightarrow 0$).*

In particular:

1. $\mathcal{T}_{\mathcal{R}^c, \lambda}$ has asymptotic exponential law w.r.t. $\mu_{\mathcal{R}}$, with rate ϕ_λ^*
2. ϕ_λ^* satisfied sharp asymptotics expressed in term of capacity

Transition time and mixing time

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From 1. and 2.

$$\mathbb{E}_{\mu_{\mathcal{R}}}(\mathcal{T}_{\mathcal{R}^c, \lambda}) = \phi_\lambda^{*-1}(1 + o(1)) = \frac{\mu(\mathcal{R})}{\text{cap}_k^\lambda(\mathcal{R}, \mathcal{R}^c)}(1 + o(1))$$

Transition time and mixing time

Moreover, the **truly escape** from \mathcal{R} is given by the time $G_{\mathcal{R}^c}(\sigma_\lambda)$, (*first excursion* $\sim \sigma_\lambda$) for $\lambda = O(S_{\mathcal{R}^c,0}^{-1})$. Indeed it holds, for all $x \in \mathcal{X}$,

$$\left\{ \begin{array}{l} \|\mathbb{P}_x(X(G_{\mathcal{R}^c}(\sigma_\lambda)) = \cdot) - \mu_{\mathcal{R}^c}\|_{\text{TV}} \leq \lambda S_{\mathcal{R}^c,0} + o(1) \\ \|\mathbb{P}_x(X(G_{\mathcal{R}^c}(\sigma_\lambda)) = \cdot) - \mu\|_{\text{TV}} \leq \mu(\mathcal{R}) + \lambda S_{\mathcal{R}^c,0} + o(1) \end{array} \right.$$

Transition time and mixing time

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$$\begin{cases} \|\mathbb{P}_x(X(G_{\mathcal{R}^c}(\sigma_\lambda)) = \cdot) - \mu_{\mathcal{R}^c}\|_{\text{TV}} \leq \lambda S_{\mathcal{R}^c,0} + o(1) \\ \|\mathbb{P}_x(X(G_{\mathcal{R}^c}(\sigma_\lambda)) = \cdot) - \mu\|_{\text{TV}} \leq \mu(\mathcal{R}) + \lambda S_{\mathcal{R}^c,0} + o(1) \end{cases}$$

THM 5. [mixing time] *If $S_{\mathcal{R}} \cdot \phi^* = o(1)$ and $S_{\mathcal{R}^c} \cdot \phi^{c*} = o(1)$ as $\varepsilon_{\mathcal{R}}, \varepsilon_{\mathcal{R}^c} \rightarrow 0$, and taking $\lambda = O(S_{\mathcal{R}^c,0}^{-1})$,*

$$\mathcal{T}_{\text{mix}} \leq \frac{4}{\gamma} \left(\frac{1 - \mu(\mathcal{R})}{1 - 2\mu(\mathcal{R})} \right) (1 + o(1))$$

Transition and mixing time of the Curie-Weiss model:

Recall that we get:

- $\mathcal{T}_{\mathcal{R}^c}$ has exponential law w.r.t. $\mu_{\mathcal{R}}$;
- $\mathbb{E}_{\mu_{\mathcal{R}}}(\mathcal{T}_{\mathcal{R}^c}) = \frac{\pi N}{\beta c(m_0)c(m_-)} e^{\beta N \Gamma_1} (1 + o(1))$;
- $\gamma^{-1} = \frac{2\pi N}{\beta c(m_0)c(m_-)} e^{\beta N \Gamma_1} (1 + o(1))$.

By Theorem 6., with no need of further computations, it holds:

(i) $\mathcal{T}_{\mathcal{R}^c, \lambda}$ has **exponential law** w.r.t. $\mu_{\mathcal{R}}$, with mean

$$\mathbb{E}_{\mu_{\mathcal{R}}}(\mathcal{T}_{\mathcal{R}^c, \lambda}) = \frac{2\pi N}{\beta c(m_0)c(m_-)} e^{\beta N \Gamma_1} (1 + o(1))$$

(ii) The **mixing time** \mathcal{T}_{mix} is bounded as

$$\gamma^{-1} \leq \mathcal{T}_{mix} \leq \frac{8\pi N}{\beta c(m_0)c(m_-)} e^{\beta N \Gamma_1} (1 + o(1)) = 4\gamma^{-1} (1 + o(1))$$

Thank you for your attention!