METASTABILITY AND SMALL EIGENVALUES
IN MARKOV CHAINS

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Abstract: In this letter we announce rigorous results that elucidate the relation between metastable states and low-lying eigenvalues in Markov chains in a much more general setting and with considerable greater precision as was so far available. This includes a sharp uncertainty principle relating all low-lying eigenvalues to mean times of metastable transitions, a relation between the support of eigenfunctions and the attractor of a metastable state, and sharp estimates on the convergence of probability distribution of the metastable transition times to the exponential distribution.

Keywords: Markov chains, metastability, eigenvalue problems, exponential distribution

I. Introduction.

The phenomenon of metastability has been a fascinating topic of non-equilibrium statistical mechanics for a long time. Currently, it has found renewed interest in the investigation of glassy systems and aging phenomena which appear to play a central role in many physical and non-physical systems.

An approach to link metastability to spectral characteristics, in particular low-lying eigenvalues and the corresponding eigenfunctions has been proposed by Gaveau and Schulman [GS]. Such an approach is appealing not only because it allows to characterize metastability in terms that are intrinsically dynamic and make no a priori reference to geometric concepts such as “free energy landscapes”, but also since it allows numerical investigations of metastable states via numerical spectral analysis (see Schütte et al. [S, SFHD] for applications to conformational dynamics of biomolecules).

Relating metastability to spectral characteristics of the Markov generator or transition matrix is in fact a rather old topic. First mathematical results go back at least as far as Wentzell [W] (see also [M] for more recent results) and Freidlin and Wentzell (see [FW]). Freidlin and Wentzell relate the eigenvalues of the transition matrix for Markov processes with exponentially small transition probabilities to exit times from “cycles”; Wentzell has a similar result for the spectral gap in the case of certain diffusion processes. All these relations are on the level of logarithmic equivalence, i.e. of the form \( \lim_{\epsilon \to 0} \epsilon \ln(\lambda^i \epsilon T^i \epsilon) = 0 \) where \( \epsilon \) is the small parameter, and \( \lambda^i , T^i \) are the eigenvalues, resp. exit times. This rather crude level of precision persists also in the more recent literature and prevents, in particular, applications to systems with unbounded numbers of metastable states which are particularly relevant for glassy systems.

In this letter we announce results that – for a large class of Markov chains – improve this situation considerably: in particular we allow for the number of metastable states to grow (with e.g. the ‘vol-
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ume'), and we give precise control of error terms for ‘finite volume’ systems. Moreover, we provide representations for all quantities concerned that are computable in terms of certain ‘escape probabilities’ that are in turn well controllable through variational representations [BEGK1]. A more detailed exposition of our results, as well as the proofs, will be given in two forthcoming papers [BEGK2, EK].

2. Stable set and metastable states.

We will consider in the sequel Markov chains $X_t$ with state space $\Gamma_N$, discrete time $t \in \mathbb{N}$, and transition matrices $P_N$. We will assume that for any fixed $N$ they are ergodic, and have a unique invariant distribution $Q_N$. We are interested in the situation when these chains exhibit “metastable” behaviour; loosely speaking, this means that the state space $G_N$ can be decomposed into subsets $S_{N,i}$ such that the typical times the process takes to go from one such set to another are much larger than the time it takes to “look like” being in equilibrium with respect to the conditional distribution $Q_N(\cdot | S_{N,i})$. Some reflection shows that this statement has considerable difficulties and cannot be interpreted literally, and that a precise definition of metastability is a rather tricky business (see e.g. the recent discussion in [BK]). We will give a precise definition that is, however, inspired by this vague consideration. The main point here is that one should make precise the two time scales we alluded to. We will take the following attitude: to look ergodic within $S_{N,i}$, the process should have at least enough time to reach the “most attractive” state within $S_{N,i}$, while at least the times to go from two such states in different metastable regions should be long compared to that time. Note that this concept is rather flexible and allows, in general, to define metastable states corresponding to different time scales.

The following definition of “stable sets” follows this ideology; however, we prefer to use certain probabilities rather than actual times as criteria, mainly because these are more readily computable. Linking them in a precise manner to times will be part of our results. We will write $\tau_x^t$, for $x \in \Gamma_N$, $I \subset \Gamma_N$, for the first non-zero time the process started at $x$ arrives at $I$.

Definition 2.1: A set $M_N \subset \Gamma_N$ will be called a set of stable points if it satisfies the following assumptions: There exist finite positive constants $a_N$, $b_N$, $c_N$, and $r_N$ such that for some sequence $\varepsilon_N$ s.t. $|\Gamma_N| \varepsilon_N \downarrow 0$, $a_N^{-1} \leq \varepsilon_N b_N$.

(i) For all $z \in \Gamma_N$,
$$P[\tau_{\bar{M}_N} < \tau_z] \geq b_N$$

(ii) For any $x \neq y \in M_N$,
$$P[\tau_y^t < \tau_x^t] \leq a_N^{-1}$$

We associate with each $x \in M_N$ its local valley
$$A(x) = \left\{ z \in \Gamma_N : P[\tau_z^t = \tau_{\bar{M}_N}] = \sup_{y \in M_N} P[\tau_y^t = \tau_{\bar{M}_N}] \right\}$$

Then
$$r_N \leq \frac{Q_N(x)}{Q_N(A(x))} \equiv R_x \geq c_N^{-1}$$

We will also write $T_{x,I} \equiv P[\tau_I^t \leq \tau_z^t]^{-1}$. An important characteristic of the sets $I \subset M_N$ is $T_I \equiv \sup_{x \in M_N} T_{x,I}$. A simplifying assumption, that will be seen to ensure sufficient “spacing” of the low lying eigenvalues is that of “genericity”, defined as follows:

\footnote{All results apply, however, also to continuous time.}
**Definition 2.2:** We say that our Markov chain is generic on the level of the set $\mathcal{M}_N$, if there exists a sequence $\epsilon_N \downarrow 0$, s.t.

(i) For all pairs $x, y \in \mathcal{M}_N$, and any set $I \subset \mathcal{M}_N \setminus \{x, y\}$ either $T_{x,I} \leq \epsilon_N T_{y,I}$ or $T_{y,I} \leq \epsilon_N T_{x,I}$.

Each of the elements of $\mathcal{M}_N$ in the generic case will then correspond indeed to a metastable state. Our first task is to identify precisely the notion of the exit time from a metastable state. To do so, we define for any $x \in \mathcal{M}_N$ the set $\mathcal{M}_N(x) = \{y \in \mathcal{M}_N : \mathbb{Q}_N(y) > \mathbb{Q}_N(x)\}$; these are the points that are even more stable than $x$. The metastable exit time, $t_x$ from $x$ is then defined as the time of the first arrival from $x$ in $\mathcal{M}_N(x)$. With this notion we can formulate our main result:

**Theorem 2.3:** Assume that $\mathcal{M}_N$ is a stable set and that the genericity assumptions are satisfied with $\epsilon_N$ such that $\tau_{NCN} \epsilon_N \downarrow 0$. Set $t_x \equiv \tau_{\mathcal{M}_N(x)}$. Then

(i) For any $x \in \mathcal{M}_N$,

$$\mathbb{E}t_x = R_x^{-1} T_{x,\mathcal{M}_N(x)}(1 + o(1)) \quad (2.5)$$

(ii) For any $x \in \mathcal{M}_N$, there exists an eigenvalue $\lambda_x$ of $1 - P_N$ that satisfies

$$\lambda_x = \frac{1}{\mathbb{E}t_x} (1 + o(1)) \quad (2.6)$$

Moreover, there exists a constant $c > 0$ such that for all $N$

$$\sigma(1 - P_N) \cup_{x \in \mathcal{M}_N} \lambda_x \subset (cb_N, 1) \quad (2.7)$$

(iii) For any $x \in \mathcal{M}_N$, for all $t > 0$,

$$\mathbb{P}[t_x > t\mathbb{E}t_x] = e^{-t(1+o(1))(1+o(1))} \quad (2.8)$$

(iv) If $\psi_x$ denotes the eigenvector of $1 - P_N$ corresponding to the eigenvalue $\lambda_x$, then

$$\psi_x(y) = \begin{cases} \mathbb{P}[\tau_y^N < \tau_{\mathcal{M}_N(x)}^N](1 + o(1)), & \text{if } \mathbb{P}[\tau_y^N < \tau_{\mathcal{M}_N(x)}^N] \geq \epsilon_N \\ O(\epsilon_N), & \text{otherwise} \end{cases} \quad (2.9)$$

**Remark:** Explicit bounds on the error terms are given [BEGK2,EK].

Let us make some additional comments on this theorem. First of all, they identification of what constitutes a metastable exit is crucial, and, in particular the fact that these processes include the *transition through the "saddle point"*, guaranteed in our case by the insistence that the process has actually arrived in $\mathcal{M}_N(x)$. Without taking this into account, the precise uncertainty principle (ii) could not hold. It is interesting to note that on the level of this theorem, the difficulties associated with the control of the passage through a saddle are not visible, and that we have the exact formula (2.5) for the mean exit time. Of course the difficulty is hidden in the quantities $T_{x,y}$ whose computation is far from trivial. However, we have shown in [BEGK1] that at least in the reversible case, using a variational representation, very precise control of such quantities can be gained in certain setting.

Somewhat less precise results can also be obtained in some non-reversible situations [EK]. Concerning our estimate on the eigenfunction, it is easy to see that [BEGK2] $\mathbb{P}[\tau_y^N < \tau_{\mathcal{M}_N(x)}^N]$ is either very close to one or very close to zero, except on a set of points whose invariant measure is extremely small. Therefore, the corresponding right-eigenfunctions $\psi_x^r(z) = \mathbb{Q}_N(z)\psi_x(z)$, are essentially proportional to the measure $\mathbb{Q}_N$ conditioned on the local valley of corresponding to $x$ (all up to errors of order $\epsilon_N$), i.e. they do indeed represent metastable measures, as suggested in [GS].
3. Some ideas of the proofs.

The first major ingredient for the proof is a representation formula for the Green’s function of the transition matrix $P_N$ in terms of certain probabilities. It implies in particular that for any $I \subset \Gamma_N$,

$$
\mathbb{E}t^I = \sum_{y \in \Gamma_N \setminus I \setminus x} Q_N(y) P[\tau_y^I < \tau_x^I] \mathbb{P}[\tau_x^I < \tau_y^I] + \frac{1}{\mathbb{P}[\tau_x^I < \tau_x^I]} \tag{3.1}
$$

This formula was first derived for the reversible case in [BEGK1]. An apparently independent derivation that also covers the non-reversible case was given recently in [GM]. This formula allows to prove in a rather simple way, (2.5). However, the realization that this formula actually arises from a representation of the Greens function makes it even more useful.

Our analysis of the spectrum of $1 - P_N$ passes through the analysis of the Laplace transforms,

$$
G_{x,I}^J(u) = \mathbb{E}e^{ut^I} \mathbb{1}_{\tau_y^I < \tau_x^I} \tag{3.2}
$$

of transition times of process that is ‘killed’ upon arrival in a set $I \subset \Gamma_N$. We write $P_N^J$ for the transition matrix of such a process, and we write $\lambda_J$ for the smallest eigenvalue of $(1 - P_N^J)$. It then turns out that all eigenvalues of $1 - P_N$ below $\lambda_J$ can be characterised as follows: Set $u(\lambda) \equiv -\ln(1 - \lambda)$. Define the $|J| \times |J|$ matrix $G_J(u)$ whose elements are

$$
\delta_{m',m} - G_{m,J}^m(u), \quad m', m \in J
$$

Then $\lambda$ is an eigenvalue of $1 - P_N$ below $\lambda_J$, if and only if

$$
\det G_J(u(\lambda)) = 0 \tag{3.3}
$$

This equation is rather easy to understand if $|J| = 1$. In this case, (3.3) becomes simply $G_{x}^m(u(\lambda)) = 1$. By a simple renewal argument, one sees that $G_x^m(u) = \frac{G^m_{x,m,x}(u)}{1 - G_{m,x}^m(u)}$. Therefore, $u(\lambda)$ defined by (3.3) is the first value at which $\sup_{x \in \Gamma_N} G_x^m(u) = +\infty$. The general formula (3.2) is somewhat less intuitive. Basically, one makes an ansatz for the eigenfunctions of $(1 - P_N)$ in terms of the Laplace transforms of the form

$$
\Psi(x) = \sum_{m \in J} \phi_m G_{m,J}^x(u) \tag{3.4}
$$

One then finds that condition (3.3) is sufficient for the ansatz to yield eigenfunctions wit $u = u(\lambda)$. Moreover, one can show that if $\lambda$ is an eigenvalue, then the eigenfunctions can be represented in this form and (3.3) must be satisfied.

To complete the proof one needs good control over the Laplace transforms; this is partly provided again by the representation of the Green’s function, complemented by lower bounds on eigenvalues $\lambda_J$ obtained from a Donsker-Varadhan [DV] argument. The actual proofs are rather involved and must be left to the longer publications [BEGK2, EK]. Let us finally mention that the good control over the spectrum of $(1 - P_N)$ allows in turn a very good control of the analytic properties of the Laplace transforms which allow in turn the sharp estimates on the probability distribution of metastable transition times stated under (iv).

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