

# Exponential decay of correlation for the Stochastic Process associated to the Entropy Penalized Method

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## Abstract

In this paper we present an upper bound for the decay of correlation for the stationary stochastic process associated with the Entropy Penalized Method. Let  $L(x, v) : \mathbb{T}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  be a  $C^1$  Lagrangian of the form

$$L(x, v) = \frac{1}{2} |v|^2 - U(x) + \langle P, v \rangle.$$

We point out that we do not assume more differentiability of  $L$  according to the dimension of the torus  $\mathbb{T}^n$ .

For each value of  $\epsilon$  and  $h$ , consider the operator

$$\mathcal{G}[\phi](x) := -\epsilon h \ln \left[ \int_{\mathbb{R}^N} e^{-\frac{hL(x,v)+\phi(x+hv)}{\epsilon h}} dv \right],$$

as well as the reversed operator

$$\bar{\mathcal{G}}[\phi](x) := -\epsilon h \ln \left[ \int_{\mathbb{R}^N} e^{-\frac{hL(x+hv,-v)+\phi(x+hv)}{\epsilon h}} dv \right],$$

both acting on continuous functions  $\phi : \mathbb{T}^n \rightarrow \mathbb{R}$ . Denote by  $\phi_{\epsilon,h}$  the solution of  $\mathcal{G}[\phi_{\epsilon,h}] = \phi_{\epsilon,h} + \lambda_{\epsilon,h}$ , and by  $\bar{\phi}_{\epsilon,h}$  the solution of  $\bar{\mathcal{G}}[\bar{\phi}_{\epsilon,h}] = \bar{\phi}_{\epsilon,h} + \lambda_{\epsilon,h}$ . Let  $\theta_{\epsilon,h}(x) = e^{-\frac{\bar{\phi}_{\epsilon,h}(x)+\phi_{\epsilon,h}(x)}{\epsilon h}}$ . From [GV], it is known that

$$\mu_{\epsilon,h}(x, v) = \theta_{\epsilon,h}(x) \gamma_{\epsilon,h}(x, v) = \theta_{\epsilon,h}(x) e^{-\frac{hL(x,v)+\phi_{\epsilon,h}(x+hv)-\bar{\phi}_{\epsilon,h}(x)-\lambda_{\epsilon,h}}{\epsilon h}},$$

is a solution to the entropy penalized problem:  $\min\{\int_{\mathbb{T}^n \times \mathbb{R}^N} L(x, v) d\mu(x, v) + \epsilon S[\mu]\}$ , where the entropy  $S$  is given by

$$S[\mu] = \int_{\mathbb{T}^n \times \mathbb{R}^N} \mu(x, v) \ln \frac{\mu(x, v)}{\int_{\mathbb{R}^N} \mu(x, w) dw} dx dv,$$

and the minimization is made over all holonomic probability densities on  $\mathbb{T}^n \times \mathbb{R}^n$ , that is probabilities that satisfy  $\int \varphi(x+v) - \varphi(x) \mu(x, v) dx dv = 0$ , for all  $\varphi \in C^1(\mathbb{T}^n)$ . The density  $\gamma_{\epsilon,h}(x, v)$  defines a Markovian transition kernel on  $(\mathbb{T}^n)^{\mathbb{N}}$ . The invariant initial density in  $\mathbb{T}^n$  is  $\theta_{\epsilon,h}(x)$ . In order to analyze the decay of correlation for this process we show that the operator  $\mathcal{L}(\varphi)(x) = \int e^{-\frac{hL(x,v)}{\epsilon}} \varphi(x+hv) dv$ , has a maximal eigenvalue isolated from the rest of the spectrum.

# 1 Definitions and the set up of the problem

Let  $\mathbb{T}^n$  be the  $n$ -dimensional torus. In this paper we assume that the Lagrangian,  $L(x, v) : \mathbb{T}^n \times \mathbb{R}^N \rightarrow \mathbb{R}$  has the form

$$L(x, v) = \frac{1}{2} |v|^2 - U(x) + \langle P, v \rangle,$$

where  $U \in C^1(\mathbb{T}^n)$ , and  $P \in \mathbb{R}^n$  is constant.

We consider here the discrete time Aubry-Mather problem [Gom] and the Entropy Penalized Mather method which provides a way to obtain approximations by continuous densities of the Aubry-Mather measure. We refer the reader to [Gom] and the last section of [GLM] for some of the main properties of Aubry-Mather measures, subactions, Peierl's barrier, etc...

The Entropy Penalized Mather problem (see [GV] for general properties of this problem) can be used to approximate Mather measures [CI] by means of absolutely continuous densities  $\mu_{\epsilon, h}(x)$ , when  $\epsilon, h \rightarrow 0$ , both in the continuous case or in the discrete case. In [GLM] it is presented a Large Deviation principle associated to this procedure. We briefly mention some definitions and results.

Consider, for each value of  $\epsilon$  and  $h$ , the operators acting on continuous functions  $\phi$ :

$$\mathcal{G}[\phi](x) := -\epsilon h \ln \left[ \int_{\mathbb{R}^N} e^{-\frac{hL(x, v) + \phi(x + hv)}{\epsilon h}} dv \right],$$

and

$$\bar{\mathcal{G}}[\phi](x) := -\epsilon h \ln \left[ \int_{\mathbb{R}^N} e^{-\frac{hL(x - hv, v) + \phi(x - hv)}{\epsilon h}} dv \right].$$

Denote by  $\phi_{\epsilon, h}$  the solution of  $\mathcal{G}[\phi_{\epsilon, h}] = \phi_{\epsilon, h} + \lambda_{\epsilon, h}$ , and by  $\bar{\phi}_{\epsilon, h}$  the solution of  $\bar{\mathcal{G}}[\bar{\phi}_{\epsilon, h}] = \bar{\phi}_{\epsilon, h} + \lambda_{\epsilon, h}$ . Let

$$\theta_{\epsilon, h}(x) = e^{-\frac{\bar{\phi}_{\epsilon, h}(x) + \phi_{\epsilon, h}(x)}{\epsilon h}}$$

By adding a suitable constant to  $\phi_{\epsilon, h}$  or  $\bar{\phi}_{\epsilon, h}$ , we can assume that  $\theta_{\epsilon, h}(x)$  is a probability density on  $\mathbb{T}^N$ . From D. Gomes and E. Valdinoci, it is known that the probability measure on  $\mathbb{T}^N \times \mathbb{R}^N$

$$\mu_{\epsilon, h}(x, v) = \theta_{\epsilon, h}(x) e^{-\frac{hL(x, v) + \phi_{\epsilon, h}(x + hv) - \phi_{\epsilon, h}(x) - \lambda_{\epsilon, h}}{\epsilon h}},$$

is a solution to the entropy penalized Mather problem:

$$\min_{\mathcal{M}_h} \left\{ \int_{\mathbb{T}^N \times \mathbb{R}^N} L(x, v) d\mu(x, v) + \epsilon S[\mu] \right\},$$

where the entropy  $S$  is given by

$$S[\mu] = \int_{\mathbb{T}^N \times \mathbb{R}^N} \mu(x, v) \ln \frac{\mu(x, v)}{\int_{\mathbb{R}^N} \mu(x, w) dw} dx dv,$$

and

$$\mathcal{M}_h := \left\{ \mu \in \mathcal{M}; \int_{\mathbb{T}^N \times \mathbb{R}^N} \varphi(x + hv) - \varphi(x) d\mu = 0, \forall \varphi \in C(\mathbb{T}^N) \right\}. \quad (1)$$

Here  $\mathcal{M}$  denotes the set of probability densities on  $\mathbb{T}^N \times \mathbb{R}^N$  and we will call  $\mu \in \mathcal{M}_h$  a holonomic probability measure.

We will be interested in measures that minimize the functional below (under the holonomic constrain)

$$\int_{\mathbb{T}^N \times \mathbb{R}^N} L(x, v) d\mu(x, v) + \epsilon S[\mu]. \quad (2)$$

Note that, for a probability  $\mu(x, v)$  the value

$$-S[\mu] = - \int_{\mathbb{T}^N \times \mathbb{R}^N} \mu(x, v) \ln \frac{\mu(x, v)}{\int_{\mathbb{R}^N} \mu(x, w) dw} dx dv$$

is not necessarily positive.

This is the entropy penalized version of the discrete time Aubry-Mather problem, see [Gom], where we look for probability measures  $\mu \in \mathcal{M}_h$  that minimize the action

$$\int_{\mathbb{T}^N \times \mathbb{R}^N} L(x, v) d\mu(x, v) \quad (3)$$

**Definition 1:** The forward (non-normalized) Perron operator  $\mathcal{L}$  is defined

$$x \rightarrow \varphi(x) \Rightarrow x \rightarrow \mathcal{L}(\varphi)(x) = \int e^{-\frac{L(x, v)}{\epsilon}} \varphi(x + hv) dv,$$

In [GV] it is shown that  $\mathcal{L}$  has a unique eigenfunction  $e^{-\frac{\phi_{\epsilon, h}}{h \epsilon}}$  with eigenvalue  $e^{-\frac{\lambda_{\epsilon, h}}{h \epsilon}}$

**Definition 2:** The backward operator  $\mathcal{N}$  is given by

$$x \rightarrow \varphi(x) \Rightarrow x \rightarrow \mathcal{N}(\varphi)(x) = \int e^{-\frac{L(x-hv, v)}{\epsilon}} \varphi(x - hv) dv,$$

In [GV] it is shown that  $\mathcal{N}$  has a unique eigenfunction  $e^{-\frac{\bar{\phi}_{\epsilon, h}}{h \epsilon}}$  with eigenvalue  $e^{-\frac{\lambda_{\epsilon, h}}{h \epsilon}}$

**Definition 3:** The operator

$$g(x) \rightarrow \mathcal{F}(g)(x) = \int e^{-\frac{hL(x, v) + \phi_{\epsilon, h}(x+hv) - \phi_{\epsilon, h}(x) - \lambda_{\epsilon, h}}{\epsilon h}} g(x + hv) dv,$$

is the normalized forward Perron operator.

From [GV] we have that given a continuous function  $g : \mathbb{T}^n \rightarrow \mathbb{R}$ , then  $\mathcal{F}^m(g)$  converges to the unique eigenfunction  $k$  as  $m \rightarrow \infty$ . We show in this paper that for  $\epsilon$  and  $h$  fixed, the convergence is exponentially fast.

Our notation:

$$\theta = \theta_{\epsilon, h}(x) = e^{-\frac{\bar{\phi}_{\epsilon, h}(x) + \phi_{\epsilon, h}(x)}{\epsilon h}},$$

$$\gamma(x, v) = \gamma_{\epsilon, h}(x, v) = e^{-\frac{hL(x, v) + \phi_{\epsilon, h}(x+hv) - \phi_{\epsilon, h}(x) - \lambda_{\epsilon, h}}{\epsilon h}},$$

in such way that  $\mu_{\epsilon, h} = \theta_{\epsilon, h}(x) \gamma_{\epsilon, h}(x, v)$ .

## 2 Reversed Markov Process and Adjoint Operator

In this section we define the reversed Markov process and compute the adjoint of  $\mathcal{F}$  in  $\mathcal{L}^2(\theta)$ . We assume  $h = 1$  from now on.

We can consider the stationary forward Markovian process  $X_n$  according to the initial probability  $\theta(x)$  and transition  $\gamma(x, v)$ . For example

$$P(X_0 \in A_0) = \int_{x \in \mathbb{T}^n \cap A_0} \theta(x) dx,$$

$$P(X_0 \in A_0, X_1 \in A_1) = \int_{x \in \mathbb{T}^n \cap A_0, (x+v) \in A_1} \theta(x) \gamma(x, v) dx dv,$$

and so on. Define the backward transfer operator  $\mathcal{F}^*$  acting on continuous functions  $f(x)$  by

$$\mathcal{F}^*(f)(x) = \int \frac{\theta(x-v) \gamma(x-v, v)}{\theta(x)} f(x-v) dv.$$

The backward transition kernel is given by

$$Q(x, v) = \frac{\theta(x-v) \gamma(x-v, v)}{\theta(x)}.$$

The fact that for any  $x$  we have  $\int Q(x, v) dv = 1$  follows from Theorem 32 in [GV]. We will show in Corollary 1 that  $\theta$  is an invariant measure for the process with transition kernel  $Q$ , more precisely, that

$$\int g d\theta = \int \mathcal{F}^*(g) d\theta,$$

for any  $g \in \mathcal{L}^2(d\theta)$ .

**Theorem 1.**  $\mathcal{F}^*$  is the adjoint of  $\mathcal{F}$  in  $\mathcal{L}^2(\theta)$ , that is for all  $f, g \in \mathcal{L}^2(\theta)$  then

$$\int f(x) \mathcal{F}g(x) \theta(x) dx = \int g(x) \mathcal{F}^*f(x) \theta(x) dx.$$

*Proof.* Consider  $f, g \in \mathcal{L}^2(\theta)$ , then

$$\begin{aligned} \int g(x) [\mathcal{F}^*(f)(x)] \theta(x) dx &= \int g(x) \left[ \int \frac{\theta(x-v) \gamma(x-v, v)}{\theta(x)} f(x-v) dv \right] \theta(x) dx \\ &= \int g(x) \left[ \int \theta(x-v) \gamma(x-v, v) f(x-v) dv \right] dx \\ &= \int \left[ \int [g(x) \theta(x-v) \gamma(x-v, v) f(x-v)] dx \right] dv \\ &= \int \left[ \int g(x+v) \theta(x) \gamma(x, v) f(x) dx \right] dv \\ &= \int f(x) \left[ \int \gamma(x, v) g(x+v) dv \right] \theta(x) dx \\ &= \int f(x) \left[ \int e^{-\frac{L(x, v) + \phi_{\epsilon, 1}(x+v) - \phi_{\epsilon, 1}(x) - \lambda_{\epsilon, 1}}{\epsilon}} g(x+v) dv \right] \theta(x) dx \\ &= \int f(x) [\mathcal{F}(g)(x)] \theta(x) dx, \end{aligned}$$

where we use above the change of coordinates  $x \rightarrow x - v$  and the fact that  $\mu$  is holonomic.  $\square$

**Corollary 1.** Consider the inner product  $\langle \cdot, \cdot \rangle$  in  $\mathcal{L}^2(\theta)$ . Then  $\mathcal{F}$  leaves invariant the orthogonal space to the constant functions:  $\{g \mid \langle g, 1 \rangle = \int g \, 1 \, d\theta = 0\}$ . Furthermore

$$\int g \, d\theta = \int \mathcal{F}^*(g) \, d\theta.$$

*Proof.* Note that  $\mathcal{F}(1) = 1$ , therefore

$$\int g \, 1 \, d\theta = \int g \mathcal{F}(1) \, d\theta = \int \mathcal{F}^*(g) \, d\theta.$$

Thus if  $\int g \, 1 \, d\theta = 0$  it follows  $\int \mathcal{F}^*(g) \, d\theta = 0$ .  $\square$

### 3 Spectral gap, exponential convergence and decay of correlations

From [GV] it is known that  $\mathcal{L}$  has a unique (normalized) eigenfunction  $e^{-\frac{\phi_{\epsilon,h}}{h\epsilon}}$  corresponding to the largest eigenvalue  $e^{-\frac{\lambda_{\epsilon,h}}{h\epsilon}}$ , in the next theorem we prove that this eigenvalue is separated from the rest of the spectrum.

**Theorem 2.** The largest eigenvalue of  $\mathcal{L}$  is at a positive distance from the rest of the spectrum.

*Proof.* We will prove the result for the normalized operator

$$g(x) \rightarrow \mathcal{F}(g)(x) = \int e^{-\frac{hL(x,v) + \phi_{\epsilon,h}(x+hv) - \phi_{\epsilon,h}(x) - \lambda_{\epsilon,h}}{\epsilon h}} g(x + hv) \, dv.$$

Recall from [GV] that the functions  $\phi_{\epsilon,h}(x)$  and  $\bar{\phi}_{\epsilon,h}(x)$  are differentiable. In this way we consider a new Lagrangian (adding  $\phi_{\epsilon,h}(x + hv) - \phi_{\epsilon,h}(x) - \lambda_{\epsilon,h}$ ) in such way  $\mathcal{L} = \mathcal{F}$ . We also assume  $\epsilon = 1$  and  $h = 1$  from now on.

Therefore,

$$g(x) \rightarrow \mathcal{F}(g)(x) = \int e^{-L(x,v)} g(x + v) \, dv,$$

the eigenvalue is 1, and, by the results in [GV], the corresponding eigenspace is one-dimensional and is generated by the constant functions.

Suppose there exist a sequence of  $f_p \in \mathcal{L}^2(\theta)$ ,  $p \in \mathbb{N}$ . such that

$$\mathcal{F}(f_p) = \lambda_p(f_p),$$

$\langle f_p, 1 \rangle = 0$ ,  $\lambda_p \rightarrow 1$  and  $\|f_p\| = 1$ . If the operator is compact, then the theorem follows from the classical argument: through a subsequence  $f_p \rightarrow f$ , and since  $\lambda_p \rightarrow 1$  we have  $\mathcal{F}(f) = f$ . Furthermore, since  $\langle f_p, 1 \rangle = 0$ , it follows  $\langle f, 1 \rangle = 0$ , which is a contradiction. Therefore we proceed to establish the compactness of the operator  $\mathcal{F}$ .

To establish compactness, consider  $g \in \mathcal{L}^2(\theta)$ . We claim that  $f = \mathcal{F}(g)$  is in the Sobolev space  $\mathcal{H}^1$  (see [E] for definition and properties). Indeed, for a fixed  $x$ , we will compute the derivative of  $f$ . Integrating by parts we have

$$\begin{aligned} \frac{d}{dx} f(x) &= \frac{d}{dx} (\mathcal{F}(g)(x)) \\ &= \int \left( \left[ \frac{d}{dx} g(x+v) \right] e^{-L(x,v)} - L(x,v) \left[ \frac{d}{dx} e^{-L(x,v)} \right] g(x+v) \right) dv \\ &= \int \left( \left[ \frac{d}{dv} g(x+v) \right] e^{-L(x,v)} - L(x,v) \left[ \frac{d}{dx} e^{-L(x,v)} \right] g(x+v) \right) dv \\ &= \int \left( \left[ \frac{d}{dv} e^{-L(x,v)} \right] g(x+v) - L(x,v) \left[ \frac{d}{dx} e^{-L(x,v)} \right] g(x+v) \right) dv \\ &= \int \left( \left[ \frac{d}{dv} e^{-L(x,v)} \right] - L(x,v) \left[ \frac{d}{dx} e^{-L(x,v)} \right] \right) g(x+v) dv. \end{aligned}$$

From the hypothesis about  $L$ , if  $g \in \mathcal{L}^2(\theta)$ , then indeed  $\frac{d}{dx} f$  is also in  $\mathcal{L}^2(\theta)$  (with the above derivative).

Note that, for  $v$  uniformly in a bounded set

$$\left\| \frac{d}{dx} f \right\|_2 \leq \left\| \frac{d}{dx} f \right\|_\infty \leq \left\| \left[ \frac{d}{dv} e^{-L(x,v)} \right] - L(x,v) \left[ \frac{d}{dx} e^{-L(x,v)} \right] \right\|_2 \|g\|_2.$$

Therefore,  $f$  is in the Sobolev space  $\mathcal{H}^1$ .

By iterating the procedure described above, we have that

$$g_j = \mathcal{F}^j(g) \in \mathcal{H}^j.$$

It is known that if  $j > \frac{n}{2}$ , where  $n$  is the dimension of the torus  $\mathbb{T}^n$ , then  $g_j$  is continuous Hölder continuous[E]. Thus the operator  $\mathcal{F}$  is compact and  $g_j$  is differentiable for a much more larger  $j$ . From the reasoning described before,  $f_p \rightarrow f$ , and  $\mathcal{F}(f) = f$ ,  $\langle f, 1 \rangle = 0$  and  $f$  is differentiable. It is easy to see that the modulus of concavity of  $f$  is bounded (the iteration by  $\mathcal{F}$  does not decrease it). We can add a constant to  $f$  and by linearity of  $\mathcal{F}$  we also get a new fixed point for  $\mathcal{F}$  (note that  $\mathcal{F}(1) = 1$ ). Therefore, we can assume  $f = e^{-g}$  for some  $g$ .

In this way, we obtain a contradiction with the uniqueness in Theorem 26 in[GV].  $\square$

Suppose  $\int g(x) \theta(x) dx = 0$ . For  $\epsilon, h$  fixed, then it follows from above that  $\mathcal{F}^m(g) \rightarrow 0$  with exponential velocity (according to the spectral gap).

Consider the backward stationary Markov process  $Y_n$  according to the transition  $Q(x, v)$  and initial probability  $\theta$  as above.

**Theorem 3.** Given  $f(x), g(x)$  with  $\int f(x) \theta(x) dx = \int g(x) \theta(x) dx = 0$ , it follows

$$\int g(Y_0) f(Y_n) dP \rightarrow 0,$$

with exponential velocity.

*Proof.* Note that

$$\begin{aligned} \int g(Y_0) f(Y_1) dP &= \int g(x) \left( \int Q(x, v) f(x - v) dv \right) \theta(x) dx = \\ &= \int g(x) [\mathcal{F}^*(f)(x)] \theta(x) dx = \int f(x) [\mathcal{F}(g)(x)] \theta(x) dx. \end{aligned}$$

In the same way, for any  $n$

$$\int g(Y_0) f(Y_n) dP = \int f(x) [\mathcal{F}^n(g)(x)] \theta(x) dx.$$

The exponential decay of correlation follows from this.  $\square$

**Theorem 4.** Let  $f(x), g(x) \in \mathcal{L}^2(\theta)$  be such that  $\int f(x)\theta(x)dx = \int g(x)\theta(x)dx = 0$ . Then

$$\int g(X_0) f(X_n) dP \rightarrow 0,$$

with exponential velocity.

*Proof.* Now, for analyzing the decay of the forward system,  $X_n$ , with transition  $\gamma(x, v)$ , we have to consider the backwark operator  $\mathcal{F}^*$ , use the fact that its exponential convergent, that is  $(\mathcal{F}^*)^n(g) \rightarrow 0$ , if  $\int g(x)\theta(x)dx = 0$ , and the result follows in the same way.  $\square$

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