Introduction to ergodic optimization

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Dynamique collective, systèmes couplés, et applications en biologie/écologie

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- I. Introduction

- II. Additive ergodic optimization on hyperbolic spaces
- III. Zero temperature limit in thermodynamic formalism
- IV. Discrete Aubry-Mather and Frenkel-Kontorova model
- V. Contreras genericity of periodic orbits
- VI. Towards multiplicative ergodic optimization

I. Introduction

• Additive ergodic optimization

- Hyperbolic dynamical system and SFT
- Minimizing measures and Gibbs measures
- Mañé conjecture for SFT
- Frenkel-Kontorova model
- Linear switched systems

Introduction : Additive ergodic optimization

Definition

• We consider a (discrete time) topological dynamical system

(X, f) compact, $f: X \to X$ continuous

• We consider also a continuous observable

 $\phi: X \to \mathbb{R}$, continuous

• The Birkhoff average along a finite orbit

$$A_n[\phi](x) := \frac{1}{n} \sum_{i=0}^{n-1} \phi \circ f^i(x)$$

• The ergodic minimizing value of ϕ

$$\bar{\phi} := \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{i=0}^{n-1} \phi \circ f^i(x)$$

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Introduction : Additive ergodic optimization

Questions

• How to compute the ergodic minimizing value?

$$\bar{\phi} := \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{i=0}^{n-1} \phi \circ f^i(x)$$

Remark : $\min_X(\phi) \le \bar{\phi} \le \max_X(\phi)$

• Is there a notion of optimal trajectory? A possible definition (forward optimality) coul be

$$\sup_{n\geq 1} \Big|\sum_{i=0}^{n-1} \left(\phi - \bar{\phi}\right) \circ f^i(x)\Big| = \sup_{n\geq 1} \Big|\sum_{i=0}^{n-1} \phi \circ f^i(x) - n\bar{\phi}\Big| < +\infty$$

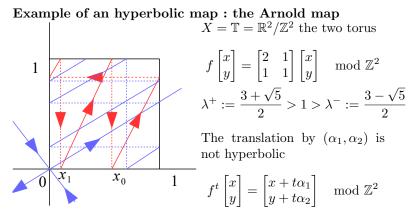
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I. Introduction

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Introduction : Hyperbolic dynamical system and SFT



Remark A C^1 perturbation of the Arnold map is hyperbolic. The class of hyperbolic maps is relatively large

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Introduction : Hyperbolic dynamical system and SFT

Another example of an hyperbolic map

Directed graph G = (V, E), $V = \{1, 2, 3\}$ $E = \{1 \rightarrow 1, 1 \rightarrow 2, 2 \rightarrow 2, \ldots\}$ $M = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 1 \\ 1 & 0 & 1 \end{bmatrix}$ The subshift of finite type **SFT** $X := \{x = (x_k)_{k \in \mathbb{Z}} : x_k \in V, x_k \rightarrow x_{k+1}\}$

Remark In fact the Arnold map and the SFT are very similar dynamics : they are both hyperbolic

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I. Introduction

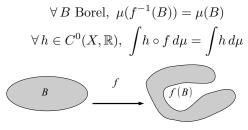
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Introduction : Minimizing and Gibbs measures

We consider a topological dynamical system (X, f) and and a continuous observable $\phi : X \to \mathbb{R}$.

Definition

• An invariant measure μ is a probability measure on X such that



Remark An hyperbolic system has many invariant measures. For instance the Arnold map preserves the normalized Lebesgue measure on \mathbb{T}^2

$$A := \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \quad \det(A) = 1 \quad \int h \circ f \operatorname{Jac} d\operatorname{Leb} = \int h \, d\operatorname{Leb}$$

(change of variable)

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Introduction : Minimizing and Gibbs measures

Recall The ergodic minimizing value

$$\bar{\phi} := \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{i=0}^{n-1} \phi \circ f^i(x)$$

Proposition We will see soon

$$\bar{\phi} = \min\left\{\int \phi \, d\mu : \mu \text{ is an invariant mesure }\right\}$$

Definition

• A minimizing measure is an invariant measure satisfying

$$\int \phi \, d\mu = \bar{\phi}$$

• The Mather set is the compact invariant set

$$Mather(\phi) := \bigcup \left\{ supp(\mu) : \mu \text{ is a minimizing measure } \right\}$$

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Introduction : Minimizing and Gibbs measures

Definition A Gibbs measure at temperature β^{-1} for the observable $\phi: X \to \mathbb{R}$ is an invariant measure that gives a specific mass to cylinders of size n.

• A cylinder of size n is

$$B_n(x,\epsilon) := \left\{ y \in X : d(f^k(x), f^k(y)) < \epsilon, \ \forall k \in \llbracket 0, n-1 \rrbracket \right\}$$

• the Gibbs measure at inverse temperature β

$$\mu_{\beta}[B_n(x,\epsilon)] \asymp \frac{1}{Z(n,\beta)} \exp\left(-\beta \sum_{k=0}^{n-1} \phi \circ f^k(x)\right)$$

• $Z(n,\beta) := \exp(-n\beta \bar{\phi}_{\beta})$ is a normalizing factor

$$-\beta\bar{\phi}_{\beta} := \lim_{n \to +\infty} \inf_{E_n: \text{ covering }} \frac{1}{n} \log\Big(\sum_{x \in E_n} \exp\big(-\beta\sum_{k=0}^{n-1} \phi \circ f^k(x)\big)\Big)$$

Remark μ_{β} gives a larger mass to configurations with low energy

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Introduction : Minimizing and Gibbs measures

Question What is the relationship between minimizing measures and Gibbs measures?

Theorem We will see that, by freezing an hyperbolic system, $\beta \to +\infty$, the Gibbs measure μ_{β} associated to a short range observable tends to a "selected" minimizing measure with maximal entropy among all minimizing measures.

Observation Some minimizing measures corresponds to "ground states" (limits up to a subsequence of Gibbs measures),; other minimizing measures have no relationship with Gibbs measures. For non short range observables, the phenomenon of "chaotic convergence" described by van Enter takes place.

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Introduction : Mañé conjecture for SFT

Recall The Mather set

Mather :=
$$\bigcup \left\{ supp(\mu) : \mu \text{ is a minimizing measure} \right\}$$

Question What is the structure of the Mather set? Is it small and reduced to a periodic orbit? Is it a set with large complexity (or entropy)? Could it be the whole set X?

Mañé Conjecture For any hyperbolic dynamical system, the Mather set is reduced to a periodic orbit for generic smooth observable.

Contreras Theorem For every subshift of finite type, for every Hölder observable $\phi : X \to \mathbb{R}$, for every perturbation $\epsilon > 0$, there exists a periodic orbit \mathcal{O}_{ϵ} such that

$$\psi := \phi + \epsilon d(\cdot, \mathcal{O}_{\epsilon})$$

has a unique minimizing measure, which is the measure supported by $\ensuremath{\mathbb{O}}$

$$\delta_{\mathcal{O}} = \frac{1}{\operatorname{card}(\mathcal{O}_{\epsilon})} \sum_{p \in \mathcal{O}_{\epsilon}} \delta_{p}$$

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Introduction : Mañé conjecture for SFT

Obvious example Every compact invariant set $\Lambda \subset X$ can play the role of a Mather set

$$\phi(x) := d(x, \Lambda) \quad \bar{\phi} = 0, \quad \mu \text{ is minimizing } \Leftrightarrow \operatorname{supp}(\mu) \subset \Lambda$$

Another example Assume the Mather set satisfies the "subordination principle" and contains a periodic orbit O then

$$\psi := \phi + \epsilon d(x, \mathcal{O})$$

has a unique minimizing measure supported in ${\mathcal O}$

Proof

2 The Mather set satisfies the subordination principle : every measure supported in the Mather set is minimizing

- **8** $\delta_{\mathbb{O}}$ is minimizing : $\bar{\psi} \leq \int \psi \, d\delta_{\mathbb{O}} = \int \phi \, d\mu_{\mathbb{O}} = \bar{\phi}$
- (4) if μ is ψ -minimizing $\int \psi \, d\mu = \bar{\psi} = \bar{\phi} \leq \int \phi \, d\mu$

$$\epsilon \int d(\cdot, \mathcal{O}) \, d\mu = \int (\psi - \phi) \, d\mu \le 0 \Rightarrow \operatorname{supp}(\mu) \subset \mathcal{O}$$

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Introduction : Frenkel-Kontorova model

Simplification The manifold is the d-torus $M = \mathbb{T}^d$, the tangent space is $TM = \mathbb{T}^d \times \mathbb{R}^d$, $\forall (x, v) \in TM$, x = position, v = velocity

Definition

(1) A Tonelli Lagrangian is a function $L(x, v) : TM \to \mathbb{R}$ which is C^2 , periodic in x, and uniformly strictly convex in v

$$\exists \alpha > 0, \ \forall x \in M, \ \operatorname{Hess}(L)(x,v) := \frac{\partial^2 L}{\partial v^2}(x,v) > \alpha$$

(2) The action of a C^1 path $\gamma:[a,b]\to M$ is the quantity

$$\mathcal{A}(\gamma) := \int_{a}^{b} L(\gamma(t), \gamma'(t)) \, dt$$

(3) The Lagrangian flow is the flow on the tangent space

$$\begin{split} \Phi^t_L(x,v) &: TM \to TM, \quad \gamma_{x,v}(t) = pr^1 \circ \Phi^t_L(x,v), \\ & \frac{d}{dt} \gamma_{x,v} = pr^2 \circ \Phi^t_L(x,v) \end{split}$$

where $\gamma_{x,v}$ is a local minimizer of the action :

$$\mathcal{A}(\gamma_{x,v}) \leq \mathcal{A}(\gamma), \quad \forall \gamma: [a,b] \to \underbrace{M}_{\mathsf{CD}}, \underbrace{C^1_{\mathsf{Close}}}_{\mathsf{CD}} \underset{\mathsf{CD}}{\mathsf{Close}}$$

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Introduction : Frenkel-Kontorova model

Example $M = \mathbb{T}^d$, $TM = \mathbb{T}^d \times \mathbb{R}^d$, $U : M \to \mathbb{R}$ a C^2 periodic function, $\lambda \in \mathbb{R}^d$ a constant representing a cohomological constraint

$$L(x, v) = \frac{1}{2} ||v||^2 - U(x) - \lambda \cdot v$$

Recall The action of a C^1 path $\gamma : [a, b] \to M$ is the quantity

$$\mathcal{A}(\gamma) := \int_{a}^{b} L(\gamma(t), \gamma'(t)) \, dt, \quad \gamma(a) = x, \ \gamma(b) = y$$

Discrete Aubry-Mather A discretization in time of a Laganrgian flow. Let $\tau > 0$ be a small number

$$\mathcal{A}_{\tau}(x,y) := \tau L\left(x, \frac{y-x}{\tau}\right) - \tau U(x) - \lambda \cdot (y-x)$$

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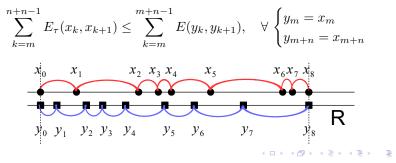
Introduction : Frenkel-Kontorova model

Frenkel-Kontorova model A discretization in time of the inverse pendulum : $d = 1, M = \mathbb{T}, \tilde{M} = \mathbb{R} \to M$ is the natural covering space

$$E_{\tau}(x,y) := \frac{1}{2\tau} |y-x|^2 + \frac{\tau K}{2\pi} (1 - \cos(2\pi x)) - \lambda(y-x)$$

 E_{τ} is called an interaction energy

Definition A minimizing configuration $(x_k)_{k \in \mathbb{Z}}, x_k \in \mathbb{R}, \forall m \in \mathbb{Z}, \forall n \ge 1$



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Introduction : Frenkel-Kontorova model

Dynamical system (Σ, σ) where Σ is the space of minimizing configurations $x = (x_k)_{k \in \mathbb{Z}}$, and $\sigma : \Sigma \to \Sigma$ is the left shift

$$\sigma(x) = y = (y_k)_{k \in \mathbb{Z}} \iff y_k = x_{k+1}, \ \forall k \in \mathbb{Z}$$

Definition The ergodic minimizing value of E, or the *effective energy*

$$\bar{E}_{\tau} = \lim_{n \to +\infty} \frac{1}{n} \inf_{x_0, x_1, \dots, x_n} \sum_{k=0}^{n-1} E(x_k, x_{k+1})$$

Proposition We will see that one can define a discrete Lagrangian dynamics $\Phi_{L,\tau}(x,v) : \mathbb{T} \times \mathbb{R} \to \mathbb{T} \times \mathbb{R}$ such that

$$\bar{E}_{\tau} = \inf \left\{ \int E(x, x + \tau v) \, d\mu(x, v) : \mu \text{ is } \Phi_{L, \tau} \text{ minimizing } \right\}$$

Remark Although $\Phi_{L,\tau}$ is not hyperbolic, a similar theory can be applied. Numerically by discretizing the space, we get back to subshift of finite type

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Introduction : Linear switched systems

Question We studied in different examples the notion of ergodic minimizing value of a scalar function $\phi : X \to \mathbb{R}$. If f is multivalued what can be said?

Definition A (discrete in time) linear switch system is a dynamical system of the form

$$v_{k+1} = A_k v_k, \ \forall k \ge 0$$

where $v_k \in \mathbb{R}^d$ represents the state of the system, $A_k \in \operatorname{Mat}(\mathbb{R}, d)$ is a square matrix, and v_{k+1} is the state at the next time. The action A_k can be chosen either by an external observer or by an automatic dynamical system (X, f)

Definition We consider a topological dynamical system (X, f), a continuous matrix function $A: X \to Mat(\mathbb{R}, d)$, and a matrix cocycle

$$A(x,n) := A \circ f^{n-1}(x) \cdots A \circ f(x)A(x)$$

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Introduction : Linear switched systems

Question One of the main problem in control theory is to stabilize a system, that is to find a trajectory $x \in X$ such that

$$||A(x,n)|| = ||A \circ f^{n-1}(x) \cdots A \circ f(x)A(x)|| \le 1$$

We are left to study the worst case, that is to compute the following characteristic of the system

Definition The maximizing singular value of a cocycle

$$\bar{\sigma}_1(A) := \lim_{n \to +\infty} \sup_{x \in X} \|A(x, n)\|^{1/r}$$

Actually we prefer to introduce the maximizing Lyapunov exponent

$$\bar{\lambda}_1 := \log(\bar{\sigma}_1(A)) = \lim_{n \to +\infty} \frac{1}{n} \sup_{x \in X} \log(\|A(x, n)\|)$$

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Introduction : Linear switched systems

Definition A cocycle of order 1 over the full shift :

- (1) a finite set of matrices $\mathcal{A} := \{M_1, \cdots, M_r\}$
- (2) the full shift $X := \mathcal{A}^{\mathbb{N}} = \{x = (A_k)_{k \ge 0} : A_k \in \mathcal{A}, \forall k \ge 0\}$ $f : X \to X$ is the left shift
- (3) the cocycle of order $1:A(x)=A_0$ if $x=(A_k)_{k\geq 0}$ $A(x,n)=A_{n-1}\cdots A_1A_0$

Example A cocycle of order 1 over a set of two matrices

$$M_1 := \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad M_2 := \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

Although the spectral radius of each matrix is

$$\rho = \lim_{n \to +\infty} \|M_1^n\|^{1/n} = \lim_{n \to +\infty} \|M_2^n\|^{1/n} = 1$$

we will see that the "generalized spectral radius" of indifferent products is

$$\lim_{n \to +\infty} \sup_{A_{n-1}, \dots, A_1, A_0} \|A_{n-1} \cdots A_1 A_0\|^{1/n} > 1$$

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II. Additive ergodic optimization on hyperbolic spaces

• Basic definitions again

- Minimal systems and Gottschalk-Hedlund
- Minimizing measures and Mather set
- An example of hyperbolic space : Subshift of finite type
- Lax-Oleinik operator and calibrated subactions
- Some extensions for Anosov systems

Additive cocycle : Basic definitions again

Definition We consider

- (1) (X, f) a topological dynamical system, X compact, $f: X \to X$ continuous
- (2) $\phi: X \to \mathbb{R}$ a continuous observable
- (3) the ergodic minimizing value of ϕ

$$\bar{\phi} := \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x)$$

Question Can we say something for the lower bound of

$$\inf_{n \ge 1} \inf_{x \in X} \left\{ \sum_{k=0}^{n-1} \phi \circ f^k(x) - n\bar{\phi} \right\}$$

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Additive cocycle : Basic definitions again

Definition A coboundary is a special observable of the form

$$\phi = u \circ f - u$$

for some continuous function $u:X\to \mathbb{R}$

An easy example Assume ϕ is a coboundary $\phi = u \circ f - u$ then

$$\bar{\phi} = 0$$
 and $\sup_{n \ge 1} \sup_{x \in X} \left| \sum_{k=0}^{n-1} \phi \circ f^k(x) - n\bar{\phi} \right| < +\infty$

Proof The Birkhoff sum can be evaluated easily

$$\begin{split} \sum_{k=0}^{n-1} \phi \circ f^k &= u \circ f^n - u \\ \sup_{x \in X} \Big| \sum_{k=0}^{n-1} \phi \circ f^k(x) \Big| &\leq 2 \|u\|_{\infty} \\ \bar{\phi} &= \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x) = 0 \end{split}$$

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II. Additive ergodic optimization on hyperbolic spaces

- Basic definitions again
- Minimal systems and Gottschalk-Hedlund
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Additive cocycle : Gottschalk-Hedlund theorem

Definition A minimal system (X, f) is a topological dynamical system so that every orbit is dense

$$\forall x \in X, \ \overline{\{f^n(x) : n \ge 0\}} = X$$

Example The hull of the Fibonacci sequence

1 the substitution : $0 \rightarrow 1, 1 \rightarrow 10$

 $0 \to 1 \to 10 \to 10.1 \to 101.10 \to 10110.101 \to 10110101.10110$

$$\omega_0 = 0, \ \omega_1 = 1, \ \omega_{n+1} = \omega_n \omega_{n-1} \quad \to \quad \omega_\infty \in \{0, 1\}^{\mathbb{N}}$$

2 the hull of the bi-infinite Fibonacci sequence $\omega_{\infty\infty}$

$$\omega_{\infty\infty} := 0^{\infty} \mid \omega_{\infty} \in \Sigma := \{0, 1\}^{\mathbb{Z}}$$
$$X := \bigcap_{n \ge 1} \overline{\{\sigma^k(\omega_{\infty\infty}) : k \ge n\}} \subseteq \Sigma$$

3 (X, σ) is a subshift of (Σ, σ) semi-conjugated to the rotation on the circle of rotation number

$$\alpha = \frac{1 + \sqrt{5}}{2} \quad \text{largest eigenvalue of} \begin{bmatrix} 0 & 1\\ 1 & 1 \end{bmatrix}$$

Additive cocycle : Gottschalk-Hedlund theorem

Theorem(Gottschalk-Hedlund) Let (X, f) be a minimal system and $\phi: X \to \mathbb{R}$ be a continuous function. Assume there exists a point $x_0 \in X$ such that

$$\sup_{n\geq 1} \Big|\sum_{k=0}^{n-1} \phi \circ f^k(x_0)\Big| < +\infty$$

Then there exists $u:X\to \mathbb{R}$ such that

$$\phi = u \circ f - u$$

(We say that ϕ is a coboundary)

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Additive cocycle : Gottschalk-Hedlund theorem

Definition A function $v: X \to \mathbb{R}$ is said to be u.s.c, upper semi continuous at $x_0 \in X$ if

$$\lim_{\epsilon \to 0} \sup_{x \in B(x_0,\epsilon)} v(x) \le v(x_0)$$

A function u is said to be l.s.c. lower semi continuous if

$$\lim_{\epsilon \to 0} \inf_{x \in B(x_0,\epsilon)} u(x) \ge u(x_0)$$

Proposition

- the supremum of a sequence of continuous functions is l.s.c.
- The infimum of a sequence of continuous functions is u.s.c.

Proposition

- v is u.s.c. $\Leftrightarrow \{v \ge \lambda\}$ is closed for every λ
- u is l.s.c. $\Leftrightarrow \{u \leq \lambda\}$ is closed for every λ

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Additive cocycle : Gottschalk-Hedlund theorem

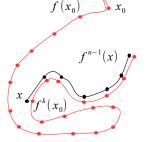
Proof of Gottschalk-Hedlund Recall we have assumed

$$R_0 := \sup_{n \ge 1} \Big| \sum_{k=0}^{n-1} \phi \circ f^k(x_0) \Big| < +\infty$$

1 We first observe that

$$\sup_{x \in X} \sup_{n \ge 1} \left| \sum_{k=0}^{n-1} \phi \circ f^k(x) \right| \le 2R_0$$

let $x \in X$, $n \ge 1$, $\epsilon > 0$ fixed. By minimality there exists $k \ge 0$



$$\sum_{i=0}^{n-1} |\phi \circ f^{i}(x) - \phi \circ f^{i+k}(x_{0})| < \epsilon$$
$$\sum_{i=0}^{n-1} \phi \circ f^{i+k}(x_{0}) = \sum_{i=0}^{n+k-1} \phi \circ f^{i}(x_{0})$$
$$-\sum_{i=0}^{k-1} \phi \circ f^{i}(x_{0})$$

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Additive cocycle : Gottschalk-Hedlund theorem

Proof of Gottschalk-Hedlund

2 We define two functions

$$u := \sup_{n \ge 1} \sum_{k=0}^{n-1} \phi \circ f^k \quad v := \inf_{n \ge 1} \sum_{k=0}^{n-1} \phi \circ f^k$$

3 u is l.s.c. v is u.s.c.

(a) the computation of $u \circ f$ and $v \circ f$ introduces a shift in the summation

$$u \circ f = \sup_{n \ge 1} \sum_{k=1}^{n} \phi \circ f^{k} \quad u \circ f + \phi = \sup_{n \ge 2} \sum_{k=0}^{n-1} \phi \circ f^{k} \le u$$
$$v \circ f = \inf_{n \ge 1} \sum_{k=1}^{n} \phi \circ f^{k} \quad v \circ f + \phi = \inf_{n \ge 2} \sum_{k=0}^{n-1} \phi \circ f^{k} \ge v$$

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Additive cocycle : Gottschalk-Hedlund theorem

Proof of Gottschalk-Hedlund

- **6** we just have proved : $u \circ f + \phi \le u$ $v \circ f + \phi \ge v$
- **6** define w := v u, then $w \circ f \ge w$
- **7** w is upper semi continuous $\rightarrow w$ attains its supremum
- **8** let x_* be a point maximizing w
- **9** then $X_* := \{x \in X : w(x) = w(x_*)\}$ is invariant by f
- () X_* is closed again by u.s.c. of w
- $I X_* = X$ by minimality $w = w(x_*), \ \forall x \in X$
- $v u = \text{const} \Rightarrow v \text{ and } u \text{ are continuous}$

$$u\circ f+\phi=u\quad v\circ f+\phi=v$$

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Introduction : Gottschalk-Hedlund theorem

Remark The assumptions in Gottschal-Hedlund implies $\bar{\phi} = 0$

$$\sup_{n\geq 1} \Big| \sum_{k=0}^{n-1} \phi \circ f^k(x_0) \Big| < +\infty \quad \Rightarrow \quad \bar{\phi} = \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x) = 0$$

Question Is the converse true?

Definition An additive cocyle is nondefective from below if there exists a constant C such that

$$\forall x \in X, \ \forall n \ge 0, \ \sum_{k=0}^{n-1} \phi \circ f(x) \ge n\bar{\phi} + C$$

Proposition If (X, f) is minimal and ϕ is continuous nondefective from below then

$$\phi = u \circ f - u + \bar{\phi}$$

for some continuous $u:X\to \mathbb{R}$

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II. Additive ergodic optimization on hyperbolic spaces

- Basic definitions again
- Minimal systems and Gottschalk-Hedlund
- Minimizing measures and Mather set
- An example of hyperbolic space : Subshift of finite type
- Lax-Oleinik operator and calibrated subactions
- Some extensions for Anosov systems

Additive cocycle : Minimizing measures and Mather set

Lemma If $(a_n)_{n\geq 0}$ is a sub additive sequence

$$a_{m+n} \le a_m + a_n, \ \forall m, n \ge 0$$

then

$$\lim_{n \to +\infty} \frac{a_n}{n} = \inf_{n \ge 1} \frac{a_n}{n}$$

Remark The following sequence $(a_n)_{n\geq 0}$ is supper additive

$$a_n := \inf_{x \in X} \sum_{k=0}^{n-1} \phi \circ f^k(x)$$

Corollary The limit in the definition of $\overline{\phi}$ exists

$$\lim_{n \to +\infty} \frac{1}{n} \inf_{x \in X} \sum_{k=0}^{n-1} \phi \circ f^k(x) = \sup_{n \ge 1} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x)$$

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Additive cocycle : Minimizing measures and Mather set

Definition We recall that a probability measure is invariant if

$$\forall h \in C^0(X, \mathbb{R}), \ \int h \circ f \, d\mu = \int h \, d\mu$$

Observation Let $\mathcal{M}(X, f)$ be the set of invariant measures

$$\int \phi \, d\mu = \int \left(\frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k\right) d\mu \ge \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k$$
$$\inf_{\mu \in \mathcal{M}(X,f)} \int \phi \, d\mu \ge \sup_{n \ge 1} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k$$

Proposition Actually

$$\inf_{\mu \in \mathcal{M}(X,f)} \int \phi \, d\mu = \sup_{n \ge 1} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x)$$

A measure realizing the infimum is called a minimizing measure

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$\label{eq:Additive cocycle} \mbox{Additive cocycle}: \mbox{\mathbf{Minimizing measures and Mather set}} \\ \mbox{\mathbf{Proof}}$

• for every $n \ge 1$, the infimum in $\inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x)$ is realized by a point x_n

2 let μ_n be the empirical measure along the trajectory

$$\mu_n := \frac{1}{n} \sum_{k=0}^{n-1} \delta_{f^k(x_n)}$$

③ by definition
$$\int \phi \, d\mu_n = \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x)$$

4 The space of probability measures is weak^{*} compact, there exists a subsequence of $(\mu_n)_{n\geq 1}$ converging to some probability measure μ . We check that μ is invariant

$$\int \phi \, d\mu = \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^k(x)$$

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Additive cocycle : Minimizing measures and Mather set

Definition We recall

Mather :=
$$\bigcup \{ \operatorname{supp}(\mu) : \mu \text{ is minimizing } \}$$

Proposition The Mather set is compact

Mather = supp(μ) for some minimizing measure μ

Question What is the structure of the Mather set? Is it a big set, a small set? Can we find on the Mather set optimal trajectories x that is

$$\sup_{n\geq 1} \Big|\sum_{k=0}^{n-1} \phi \circ f^k(x) - n\bar{\phi}\Big| < +\infty$$

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II. Additive ergodic optimization on hyperbolic spaces

- Basic definitions again
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Additive cocycle : Subshift of finite type

Definition We consider here a one-sided subshift of finite type

- $\mathcal{A} := \{1, 2, \cdots, r\}$ is a finite set of states
- M is a $r \times r$ square matrix describing the allowed transitions

 $M(i,j) \in \{0,1\} \quad M(i,j) = 1 \ \Leftrightarrow \ i \to j \text{ is an admissible transition}$

• $X = \{(x_n)_{n \ge 0} : \forall n \ge 0, x_n \in \mathcal{A}, M(x_n, x_{n+1}) = 1\}$

X is called a subshift of finite type SFT. The left shift $f: X \to X$

$$x = (x_0, x_1, x_2, \ldots) \quad \Rightarrow \quad y = f(x) = (x_1, x_2, x_3, \ldots)$$

• X equipped with the product topology is compact metrizable

$$d(x,y) = e^{-n} \Leftrightarrow x_0 = y_0, \cdots, x_{n-1} = y_{n-1} \text{ and } x_n \neq y_n$$

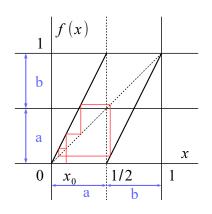
• we assume M is semi-irreducible

$$\forall i \in \mathcal{A}, \ \exists j \in \mathcal{A}, \ M(i,j) = 1$$

$$\forall j \in \mathcal{A}, \ \exists i \in \mathcal{A}, \ M(i,j) = 1$$

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Additive cocycle : Subshift of finite type



The doubling period

 $f:x\mapsto 2x \bmod 1$

is semi conjugated (up to a countable number of points) to the full shift

$$X = \{a, b\}^{\mathbb{N}}$$

Here the hyperbolicity is related to the fact that

|f'(x)| > 1

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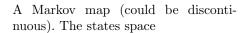
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Additive cocycle : Subshift of finite type



 $\mathcal{A} = \{a, b, c\}$

The transition matrix

$$M = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

The Markov map is semi conjugated to the SFT

$$X = \left\{ x \in \mathcal{A}^{\mathbb{N}} : M(x_k, x_{k+1}) = 1, \ \forall k \right\}$$

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Again the hyperbolicity of the Markov map is obtained because of |f'(x)| > 1. Any C^2 perturbation still remaining Markov is semi conjugated to (X, f)

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Additive cocycle : Subshift of finite type

Remark A SFT is hyperbolic in the following sense

• if $x_0 = y_0, \dots, x_{n-1} = y_{n-1}$ and $x_n \neq y_n$ then

$$\begin{aligned} d(x,y) &= e^{-n}, \quad d(f(x),f(y)) = e^{-(n-1)} = e^1 d(x,y) \\ &\Rightarrow \sigma \text{ is expanding} \end{aligned}$$

• if x and y are two configurations such that $x_0 = y_0$ and

$$\cdots x_{-3} \to x_{-2} \to x_{-1} \to x_0,$$

are preimages of x_0 then the new configurations

$$x' = (x_{-1}, x_0, x_1, \ldots) \quad y' = (x_{-1}, y_0, y_1, \ldots)$$
$$x'' = (x_{-2}, x_{-1}, x_0, x_1, \ldots) \quad y'' = (x_{-2}, x_{-1}, y_0, y_1, \ldots)$$

are contracted

$$d(x',y') = e^{-1}d(x,y) \quad d(x'',y'') = e^{-2}d(x,y)$$

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II. Additive ergodic optimization on hyperbolic spaces

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Additive cocycle : Lax-Oleinik operator

Recall The ergodic minimizing value of ϕ can be computed using measure

$$\bar{\phi} = \min\left\{\int \phi \, d\mu : \mu \text{ is an invariant measure }\right\}$$
$$Mather(\phi) := \bigcup\left\{\operatorname{supp}(\mu) : \mu \text{ is minimizing }\right\}$$

Definition An observable is nondefective from below if

$$\forall x \in X, \ \forall n \ge 0, \ \sum_{k=0}^{n-1} \phi \circ f^k(x) \ge n\bar{\phi} + C$$

Theorem(Gottschalk-Hedlund) If (X, f) is minimal and $\phi : X \to \mathbb{R}$ is continuous then : $\sup_{n \ge 1} \left| \sum_{k=0}^{n-1} \phi \circ f^k(x_0) \right| < +\infty \quad \Rightarrow \quad \phi = u \circ f - u$

Extension If (X, f) is minimal and ϕ is nondefective from below then

$$\phi = u \circ f - u + \bar{\phi}$$

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Additive cocycle : Lax-Oleinik operator

Main hypothesis The observable is Lipschitz (or Hölder)

$$\forall x, y \in X, \ x_0 = y_0, \quad |\phi(x) - \phi(y)| \le \operatorname{Lip}(\phi) d(x, y)$$

Main result If (X, f) is a SFT, if $\phi : X \to \mathbb{R}$ is Lipschitz then there exists a Lipschitz function $u : X \to \mathbb{R}$ such that

(1) $\forall x \in X, \ \phi(x) \ge u \circ f(x) - u(x) + \bar{\phi}$ (2) $\forall x \in \text{Mather}, \ \phi(x) = u \circ f(x) - u(x) + \bar{\phi}$

Definition A subaction for ϕ is a continuous function u such that

$$\forall x \in X, \ \phi(x) \ge u \circ f(x) - u(x) + \bar{\phi}$$

Corollary Every Lipschitz ϕ is non defective from below

$$\sum_{k=0}^{n-1} \phi \circ f^k(x) \ge u \circ f^n(x) - u(x) + n\bar{\phi} \ge n\bar{\phi} - 2||u||_{\infty}$$

Corollary Every trajectory of the Mather set is optimal

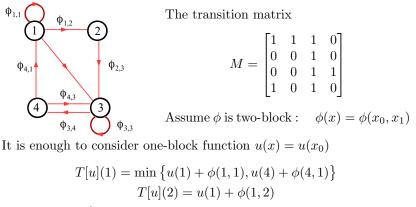
$$x \in \operatorname{Mather}(\phi) \quad \Rightarrow \quad \left|\sum_{k=0}^{n-1} \left(\phi \circ f^k(x) - \bar{\phi}\right)\right| \le 2\|u\|_{\infty}$$

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Additive cocycle : Lax-Oleinik operator

Main tool The Lax-Oleinik operator is a (nonlinear) operator acting on Lipschitz function $u: X \to \mathbb{R}$ defined by

$$T[u](y) := \min\{u(x) + \phi(x) : f(x) = y\}$$



 $T[u](3) = \min\left\{u(1) + \phi(1,3), \ u(2) + \phi_{2,3}, u(3) + \phi(3,3), \ u(4) + \phi(4,3)\right\}$

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Additive cocycle : Lax-Oleinik operator

Definition The Lax-oleinik operator $T : \operatorname{Lip}(X, \mathbb{R}) \to \operatorname{Lip}(X, \mathbb{R})$

$$T[u](y):=\min\{u(x)+\phi(x):f(x)=y\}$$

Theorem

(1) There exists a unique "additive eigenvalue" a and an (a priori non unique) "additive eigenfunction" $u \in \text{Lip}(X, \mathbb{R})$ such that

$$T[u] = u + a$$

- (2) $a = \overline{\phi}$ is the unique eigenvalue
- (3) Every eigenfunction u is a subaction

$$\phi(x) \ge u \circ f(x) - u(x) + \bar{\phi}$$

Definition An additive eigenfunction of the Lax-Oleinik operator is called a calibrated subaction

Additive cocycle : Lax-Oleinik operator

The proof uses either the Schauder theorem or a more explicit iterative scheme

Ishikawa's Algorithm(Admitted) Let \mathbb{B} be a Banach space, $\mathbb{K} \subset \mathbb{B}$ be a convex compact set, and $T : \mathbb{K} \to \mathbb{K}$ be a nonexpansive map

$$||T[u] - T[v]|| \le ||u - v||$$

Then the sequence

$$u_0 \in \mathbb{K}, \quad u_{n+1} = \frac{u_n + T[u_n]}{2}$$

converges to a fixed point.

Notation We will apply Ishikawa's algorithm to

$$\mathbb{B} := C^0(X, \mathbb{R})/\mathbb{R} \quad \text{with} \quad u \sim v \iff u - v = \text{const.}$$
$$\|\|u\|\| := \inf\{\|u + c\|_{\infty} : c \in \mathbb{R}\}$$
$$\mathbb{K}_C := \left\{ u \in \mathbb{B} : \operatorname{Lip}(u) \le C \right\} \quad \text{for some constant } C$$

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Additive cocycle : Lax-Oleinik operator

Recall The Lax-Oleinik operator : $X \subseteq \mathcal{A}^{\mathbb{N}}$, $\mathcal{A} = \{1, \ldots, r\}$

$$T[u](x_0, x_1, x_2, \ldots) = \min_{x_{-1} \in \mathcal{A}} \left\{ (u + \phi)(x_{-1}, x_0, x_1, \ldots) \right\}$$

Main observation Two points $x, y \in X$ starting at the same symbol $i_0 = x_0 = y_0 \in A$ have a common symbolic inverse branch which contracts exponentially fast

$$\begin{aligned} x_0 &= y_0 \; \Rightarrow \; \exists i_{-3} \to i_{-2} \to i_{-1} \to i_0 \\ x^{(-n)} &:= (i_{-n}, \dots, i_{-1}, x_0, x_1, \dots), \quad f^n(x^{(-n)}) = x \\ y^{(-n)} &:= (i_{-n}, \dots, i_{-1}, y_0, y_1, \dots) \\ d(x^{(-n)}, y^{(-n)}) &\leq \lambda^n d(x, y) \end{aligned}$$

for some $0 < \lambda < 1$ $(\lambda = e^{-1})$

Hyperbolicity The existence of such a contracting inverse dynamics is the main observation for the existence of u

Additive cocycle : Lax-Oleinik operator

Proof of the ergodic Lax-Oleinik's theorem

1 we recall the definition

$$T[u](y) = \min_{f(x)=y} \left(u(x) + \phi(x) \right)$$

2 T commutes with the constants : T[u + c] = T[u] + c
3 T is nonexpansive :

$$||T[u] - T[v]||_{\infty} \le ||u - v||_{\infty}$$

 $\begin{array}{lll} y \mbox{ fixed } &\Rightarrow & \exists x \mbox{ optimal}, T[v](y) = v(x) + \phi(x) \\ T[u] \mbox{ is a min } &\Rightarrow & T[u](y) \leq u(x) + \phi(x) \\ \mbox{substracting } &\Rightarrow & T[u](y) - T[v](y) \leq u(x) - v(x) \leq \|u - v\| \\ \mbox{ permuting } &\Rightarrow & |T[u](y) - T[v](y)| \leq u(x) - v(x) \leq \|u - v\| \end{array}$

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Additive cocycle : Lax-Oleinik operator

Proof of the ergodic Lax-Oleinik's theorem

(a) T preserves the set : $\begin{cases} u : \operatorname{Lip}(u) \le C \\ 0 \end{cases} C := \frac{\lambda}{1-\lambda} \operatorname{Lip}(\phi) \\ \operatorname{choose} y, y' \text{ such that } y_0 = y'_0 \\ \operatorname{optimize} T[u](y') : \exists x', \ f(x') = y' \text{ such that} \\ T[u](y') = u(x') + \phi(x') \end{cases}$

choose the same inverse branch : $\exists x, f(x) = y$ such that $d(x, x') \le \lambda d(y, y')$

by minimizing T[u](y) and substracting

$$T[u](y) \le u(x) + \phi(x)$$

$$T[u](y) - T[u](y') \le (u + \phi)(x) - (u + \phi)(x')$$

6 we use now that ϕ is Lipschitz

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Additive cocycle : Lax-Oleinik operator

Proof of the ergodic Lax-Oleinik's theorem

6 we introduce the quotient space $\mathbb{B} := C^0(X, \mathbb{R})/\mathbb{R}$ *T* acts on \mathbb{B} because *T* commutes with the constants *T* preserves the set

$$\mathbb{K} = \left\{ u \in \mathbb{B} : \operatorname{Lip}(u) \leq \frac{\lambda}{1-\lambda} \operatorname{Lip}(\phi) \right\}$$

 ${\mathbb K}$ is convex

- **7** By Ascoli's theorem \mathbb{K} is compact
- 8 by Ishikawa's theorem T admits a fixed point u in \mathbb{K} : there exists $u: X \to \mathbb{R}$ Lipschitz and $a \in \mathbb{R}$ such that

$$T[u] = u + a$$

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Additive cocycle : Lax-Oleinik operator

Proof of the ergodic Lax-Oleinik's theorem

9 We show that $a \leq \overline{\phi}$. For every $x, y \in X$

$$\begin{split} f(x) &= y \; \Rightarrow \; u(y) + a = T[u](y) \leq u(x) + \phi(x) \\ & u \circ f(x) + a \leq u(x) + \phi(x) \end{split}$$

we thus have proved that an additive eigenfunction is a subaction

$$u \circ f - u + a \le \phi$$

$$\forall x \in X, \ u \circ f^{n}(x) - u(x) + na \le \sum_{k=0}^{n-1} \phi \circ f^{k}(x)$$
$$a \le \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^{k}(x) = \bar{\phi}$$

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Additive cocycle : Lax-Oleinik operator

Proof of the ergodic Lax-Oleinik's theorem

(We show that $a \ge \overline{\phi}$. We choose arbitrarily a point $x^{(0)} \in X$. By optimality in the definition in Lax-Oleinik

$$\begin{split} u(y) + a &= T[u](y) = \min_{f(x)=y} \{ u(x) + \phi(x) \} \\ \exists x^{(-1)} \in X, \ f(x^{(-1)}) = x^{(0)}, & u(x^{(0)}) + a = u(x^{(-1)}) + \phi(x^{(-1)}) \\ \exists x^{(-2)} \in X, \ f(x^{(-2)}) = x^{(-1)}, & u(x^{(-1)}) + a = u(x^{(-2)}) + \phi(x^{(-2)}) \\ \exists x^{(-3)} \in X, \ f(x^{(-3)}) = x^{(-2)}, & u(x^{(-2)}) + a = u(x^{(-3)}) + \phi(x^{(-3)}) \\ & & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots \\ & & \sum_{k=1}^{n} \phi(x^{(-k)}) = u(x^{(0)}) - u(x^{(-n)}) + na \\ & \bar{\phi} = \lim_{n \to +\infty} \inf_{x \in X} \frac{1}{n} \sum_{k=0}^{n-1} \phi \circ f^{k}(x) \le \lim_{n \to +\infty} \frac{u(x^{(0)}) - u(x^{(-n)}) + na}{n} = a \end{split}$$

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Additive cocycle : Lax-Oleinik operator

Corollary Let (X, f) be a SFT, let ϕ be a Lipschitz function (1) there exits a Lipschitz subaction $u: X \to \mathbb{R}$

$$\forall x \in X, \ \phi(x) \ge u \circ f(x) - u(x) + \bar{\phi}$$

(2) up to a coboundary, the ergodic minimizing value is a true minimum

$$\psi := \phi - (u \circ f - u) \implies \begin{cases} \bar{\psi} = \min_X(\psi) = \bar{\phi} \\ \forall x \in X, \ \psi(x) \ge \bar{\psi} \\ \forall x \in \text{Mather}, \ \psi(x) = \bar{\psi} \end{cases}$$

Proof

for every invariant measure ∫ψ dμ = ∫φ dμ ⇒ ψ̄ = φ̄
as (ψ - φ̄) ≥ 0 and ∫(ψ - φ̄) dμ = 0 for μ minimizing
∀x ∈ supp(μ), ψ = φ̄

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Additive cocycle : Lax-Oleinik operator

Corollary Every trajectory in the Mather set is optimal

$$\forall x \in \text{Mather}, \sup_{n \ge 1} \left| \sum_{k=0}^{n-1} \phi \circ f^k(x) - n\bar{\phi} \right| < +\infty$$

Proof

1 for every minimizing measure μ $\int (\phi - \bar{\phi}) d\mu = 0$ **2** there exists a subaction $(\phi - \bar{\phi}) - (u \circ f - u) \ge 0$ **3** $\int (\phi - \bar{\phi}) - (u \circ f - u) d\mu = 0$ **4** $\phi - \bar{\phi} = u \circ f - u$ μ a.e. **5** $\phi - \bar{\phi} = u \circ f - u$ everywhere on $\operatorname{supp}(\mu)$ **6** $\left|\sum_{k=0}^{n-1} (\phi - \bar{\phi}) \circ f^k(x)\right| = |u \circ f^n(x) - u(x)| \le 2||u||_{\infty}$ on $\operatorname{supp}(\mu)$

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- I. Introduction
- II. Additive ergodic optimization on hyperbolic spaces
- III. Zero temperature limit in thermodynamic formalism
- IV. Discrete Aubry-Mather and Frenkel-Kontorova model
- V. Contreras genericity of periodic orbits
- VI. Towards multiplicative ergodic optimization

III. Zero temperature limit in thermodynamic formalism

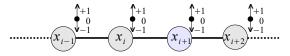
• Description of the BEG model

- Gibbs measures of a directed graph
- Ground states of a directed graph
- Zero temperature limit for a SFT
- Explicit computations for the BEG model

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Zero limit : Description of the BEG model

Description The Blume Emery Griffiths model (BEG model)



One considers a chain of atoms on a <u>lattice</u> at <u>equilibrium</u> at positive temperature that interact with their first neighbours.

- (1) Each site of the lattice hosts a unique atom
- (2) there are 3 kinds of atoms; either He⁴ with spin up or down, or an isotope He³ with no spin. Let $\mathcal{A} = \{-1, 0, 1\}$ be the 3 kinds of atoms.
- (3) a chain of atoms is an infinite sequence $x = (x_k)_{k \in \mathbb{Z}}, x_k \in \mathcal{A}$
- (4) the interaction energy is short-range and is given by an Hamiltonian : $H: \mathcal{A} \times \mathcal{A} \to \mathbb{R}$
- (5) the energy of a finite block of atoms

$$H(x_m, x_{m+1}, \dots, x_{m+n}) := \sum_{k=m}^{m+n-1} H(x_k, x_{k+1})$$

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Zero limit : Description of the BEG model

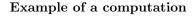
Hamiltonian in BGE $H : \mathcal{A} \times \mathcal{A} \to \mathbb{R}$ has the form

$$H(x,y) := -Jxy - Kx^2y^2 + \frac{\Delta}{2}(x^2 + y^2)$$

x, y ∈ A = {-1,0,1}
 J > 0 ⇒ spins tend to be aligned
 K > 0 ⇒ spins tend to be neighbours
 Δ > 0 ⇒ role of a chemical potential
 directed graph with transition matrix

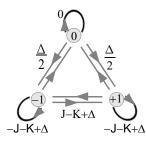
$$M = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

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$$H(0,0) = 0, \quad H(-1,1) = J - K + \Delta, \ \cdots$$

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III. Zero temperature limit in thermodynamic formalism

- Description of the BEG model
- Gibbs measures of a directed graph
- Ground states of a directed graph
- Zero temperature limit for a SFT
- Explicit computations for the BEG model

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Zero limit : Gibbs measures of a directed graph

Formal notations

- (1) $\mathcal{A} = \{1, 2, \dots, r\}$: the possible state space of the atoms
- (2) M: an $r \times r$ matrix with values in $\{0, 1\}$ called transition matrix

 $M(i,j) = 1 \iff$ a transition $i \rightarrow j$ is allowed

(3) (X,f): the bi-infinite subshift of finite type, $f:X\to X$

$$X = \left\{ x = (x_k)_{k \in \mathbb{Z}} : \forall k \in \mathbb{Z}, \ x_k \in \mathcal{A}, \ M(x_k, x_{k+1}) \right\} \subseteq \mathcal{A}^{\mathbb{Z}}$$
$$f(x) = y = (y_k)_{k \in \mathbb{Z}}, \quad \forall k \in \mathbb{Z}, \ y_k = x_{k+1}$$

(4) $H: \mathcal{A} \times \mathcal{A} \to \mathbb{R} \cup \{+\infty\}$: the Hamiltonian of the system describing the local energy between two successive atoms

$$H(i,j) = +\infty \iff M(i,j) = 0$$

(5) $\phi:X\to\mathbb{R}$: the corresponding short rang interaction on the SFT

$$\phi(x) = H(x_0, x_1)$$

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Zero limit : Gibbs measures of a directed graph

Assumption The transition matrix (or the graph) is irreducible : for every state $i, j \in \mathcal{A}$

$$\exists i = i_0 \to i_1 \to i_2 \to \dots \to i_n = j$$

Definition We introduce a weight for each transition

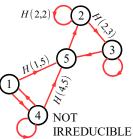
$$M_{\beta}(i,j) := \exp(-\beta H(i,j))$$

BLE which should be proportional to the probability of the occurrence of the the transition

Remark

- (1) β is supposed to be the inverse of the temperature T
- (2) M_0 is the initial transition matrix corresponding to $T = +\infty$
- (3) M_{∞} is the frozen state corresponding to T = 0

Physical Ansatz The configurations prefer transitions with low energy (\rightarrow which explains the sign $-\beta H$)



Zero limit : Gibbs measures of a directed graph

Definition A cylinder of size n is a set of configurations that have prescribed states on n consecutive sites of \mathbb{Z} . To simplify the notations, the cylinder starts at 0. If $i_0, i_1, \ldots, i_n \in \mathcal{A}$ then

$$[i_0, i_1, \dots, i_n] := \left\{ x = (x_k)_{k \in \mathbb{Z}} \in X : x_0 = i_0, \ x_1 = i_1, \dots, x_n = i_n \right\}$$

Definition The total energy of a block is

$$H(i_0, \dots, i_n) := \sum_{k=0}^{n-1} H(i_k, i_{k+1}) = \sum_{k=0}^{n-1} \phi \circ f^k(x), \quad \forall x \in [i_0, \dots, i_n]$$

Definition A Gibbs measure at temperature β^{-1} is an invariant measure of the SFT (X, f) such that

$$\mu_{\beta}([i_0,\ldots,i_n]) \asymp \exp\left(-\beta H(i_0,\ldots,i_n) + n\beta \bar{H}_{\beta}\right)$$
$$\exp(-n\beta \bar{H}_{\beta}) \asymp \sum_{\substack{[i_0,\ldots,i_n]\\\text{admissible}}} \exp\left(-\beta H(i_0,\ldots,i_n) + n\beta \bar{H}_{\beta}\right)$$

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Zero limit : Gibbs measures of a directed graph

Theorem Let (X, f) be a SFT associated to an irreducible transition matrix and $H : \mathcal{A} \times \mathcal{A} \to \mathbb{R} \cup \{+\infty\}$ be a two-step Hamiltonian. Then there exists a unique Gibbs measure at every temperature β^{-1}

Recall $M_{\beta}(i,j) = \exp(-\beta H(i,j)).$

Definition A non negative matrix $M \in Mat(\mathbb{R}^+, r)$ is said to be an irreducible matrix, if $\forall i, j \in \{1, \ldots, r\}$, there exists i_0, i_1, \ldots, i_n , with $i_0 = i$ and $j_0 = j$ such that

$$M(i_0, i_1)M_\beta(i_1, i_2)\cdots M(i_{n-1}, i_n) > 0$$

Remember $M_{\beta}(i,j) = 0 \iff H(i,j) = +\infty \iff i \not\rightarrow j$

Perron Frobenius theorem If M is a non negative irreducible matrix, then the spectral radius ρ of M is strictly positive and ρ is an eigenvalue of multiplicity 1. Moreover the eigenvector associated to 1 can be chosen to have strictly positive entries

Zero limit : Gibbs measures of a directed graph

Theorem Let (X, f) be a SFT associated to an irreducible transition matrix and $H : \mathcal{A} \times \mathcal{A} \to \mathbb{R} \cup \{+\infty\}$ be a two-step Hamiltonian. Then there exists a unique Gibbs measure at every temperature β^{-1}

Proof The Perron-Frobenius theorem tells us

- **1** let $M_{\beta}(i,j) = \exp(-\beta H(i,j))$ be an irreducible $r \times r$ matrix
- **2** let $\rho_{\beta} := \exp(-\beta \bar{H}_{\beta})$ be the largest eigenvalue
- **3** let $R_{\beta}(i)$ be the right eigenvector with strictly positive entries
- (a) let $L_{\beta}(i)$ be the left eigenvector with strictly positive entries
- **6** we normalize so that : $\sum_{i=1}^{r} L_{\beta}(i) R_{\beta}(i) = 1$

The Gibbs measure at temperature β^{-1} of a cylinder is

$$\mu_{\beta}([i_0,\ldots,i_n]) = \frac{1}{\rho_{\beta}^n} L_{\beta}(i_0) \exp\left(-\beta H(i_0,\ldots,i_n)\right) R_{\beta}(i_n)$$

We show that μ_{β} is a well defined probability on X and is invariant by the dynamics f

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Zero limit : Gibbs measures of a directed graph

Recall The Gibbs measure at temperature β^{-1} is defined by

$$\mu_{\beta}([i_0, \dots, i_n]) = \frac{1}{\rho_{\beta}^n} L_{\beta}(i_0) \Big[\prod_{k=0}^{n-1} M_{\beta}(i_k, i_{k+1}) \Big] R_{\beta}(i_n)$$

Step 1 The measure is consistent in the Kolmogorov sense

$$\sum_{j=1}^{r} \mu_{\beta}([i_0, \dots, i_n, j]) = \mu_{\beta}([i_0, \dots, i_n]) \Big[\frac{1}{\rho_{\beta}} \sum_{j=1}^{r} M_{\beta}(i_n, j) \frac{R_{\beta}(j)}{R_{\beta}(i_n)} \Big]$$
$$= \mu_{\beta}([i_0, \dots, i_n])$$

Step 2 The measure is invariant

$$\sum_{i=1}^{r} \mu_{\beta}([i, i_0, \dots, i_n]) = \left[\frac{1}{\rho_{\beta}} \sum_{i=1}^{r} \frac{L_{\beta}(i)}{L_{\beta}(i_0)} M_{\beta}(i, i_0)\right] \mu_{\beta}([i_0, \dots, i_n])$$
$$= \mu_{\beta}([i_0, \dots, i_n])$$

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III. Zero temperature limit in thermodynamic formalism

- Description of the BEG model
- Gibbs measures of a directed graph
- Ground states of a directed graph
- Zero temperature limit for a SFT
- Explicit computations for the BEG model

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Zero limit : Ground states of a directed graph

Recall The Gibbs measure of a two steps cylinder is given by

$$\mu_{\beta}([i,j)] = L_{\beta}(i) \frac{M_{\beta}(i,j)}{\rho_{\beta}} R_{\beta}(j), \qquad M_{\beta}(i,j) = \exp(-\beta H(i,j))$$

where ρ_{β} is the largest eigenvalue of M_{β}

Definition Let \bar{H}_{β} be the free energy at temperature β^{-1} defined by

$$\rho_{\beta} := \exp(-\beta \bar{H}_{\beta})$$

Question What is the behaviour of the free energy \bar{H}_{β} when the system is frozen?

Question What is the behaviour of the Gibbs measure μ_{β} when the system is frozen?

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Zero limit : Ground states of a directed graph

Proposition The free energy converges to the ergodic minimizing value $\bar{\phi}$

$$\lim_{\beta \to +\infty} \bar{H}_{\beta} = \bar{H} =: \inf_{\mu} \sum_{i=1}^{r} \sum_{j=1}^{r} H(i,j)\mu(i,j)$$

where the infimum is realized over the set of probability measures μ on $\mathcal{A} \times \mathcal{A}$ satisfying the invariance property

$$\forall i \in \mathcal{A}, \ \mu^{(1)}(i) := \sum_{k=1}^{r} \mu(i,k) = \sum_{k=1}^{r} \mu(k,i) =: \mu^{(2)}(i)$$

Theorem The Gibbs measure μ_{β} converges to a selected minimizing measure μ_{min} , that is a probability measure satisfying the previous invariance and

$$\sum_{i=1}^{r} \sum_{j=1}^{r} H(i,j) \mu_{min}(i,j) = \bar{H}$$

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Zero limit : Ground states of a directed graph

Proof of $\bar{H}_{eta} o \bar{H}$

() we recall some notations $\mathcal{A} = \{1, \ldots, r\}$

$$M_{\beta}(i,j) = \exp(-\beta H(i,j)), \quad \rho_{\beta} = exp(-\beta \bar{H}_{\beta})$$

2 we choose another left eigenvector

$$\forall j \in \mathcal{A}, \ \sum_{i=1}^{r} L_{\beta}(i) M_{\beta}(i,j) = \rho_{\beta} L_{\beta}(j), \quad \max_{i} L_{\beta}(i) = 1$$

3 we change L_{β} to an exponential form

$$L_{\beta}(i) := \exp(-\beta U_{\beta}(i)), \quad \min_{i} U_{\beta}(i) = 0$$

4 the eigenvalue problem becomes

$$\forall j \in \mathcal{A}, \quad \sum_{i=1}^{r} \exp\left(-\beta \left(H(i,j) - \bar{H}_{\beta} - (U_{\beta}(j) - U_{\beta}(i))\right)\right) = 1$$

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${\rm Zero\ limit}: {\bf Ground\ states\ of\ a\ directed\ graph}$ Proof of $ar{H}_eta o ar{H}$

6 we recall the new eigenvalue problem

$$\forall j \in \mathcal{A}, \quad \sum_{i=1}^{r} \exp\left(-\beta\left(H(i,j) - \bar{H}_{\beta} - (U_{\beta}(j) - U_{\beta}(i))\right)\right) = 1$$

6 first consequence

$$\begin{cases} \forall i \to j \in \mathcal{A}, & U_{\beta}(j) + \bar{H}_{\beta} \leq U_{\beta}(i) + H(i,j) \\ \forall j \in \mathcal{A}, \ \exists i \in \mathcal{A}, \quad \frac{\log(r)}{\beta} + U_{\beta}(j) + \bar{H}_{\beta} \geq U_{\beta}(i) + H(i,j) \end{cases}$$

 $onumber \circ ext{ second consequence, by irreducibility of the transition matrix, and the fact that there exists <math>i_0 \in \mathcal{A}$ such that $U_{\beta}(i_0) = 0$, one can find $N \geq 1$

$$0 \le \max_{j} U_{\beta}(j) \le \max_{1 \le n \le N} \max_{i=i_0 \to \dots \to i_n=j} \left(H(i_0, \dots, i_n) - n\bar{H}_{\beta} \right) < +\infty$$

 \bar{H}_{β} and $U_{\beta}(j)$ are uniformly bounded with respect to β

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Zero limit : Ground states of a directed graph

Proof of $\bar{H}_{eta} o \bar{H}$

§ \bar{H}_{β} and $U_{\beta}(j)$ are uniformly bounded with respect to β by taking a subsequence $\beta \to +\infty$

$$\lim_{\beta \to +\infty} U_{\beta}(i) = U(i), \quad \lim_{\beta \neq +\infty} \bar{H}_{\beta} = \bar{H}$$

9 we recall

$$\begin{cases} \forall i \to j \in \mathcal{A}, & U_{\beta}(j) + \bar{H}_{\beta} \leq U_{\beta}(i) + H(i,j) \\ \forall j \in \mathcal{A}, \ \exists i \in \mathcal{A}, \quad \frac{\log(r)}{\beta} + U_{\beta}(j) + \bar{H}_{\beta} \geq U_{\beta}(i) + H(i,j) \end{cases}$$

(passing to the limit $\beta \to +\infty$

$$\begin{cases} \forall i \to j \in \mathcal{A}, & U(j) + \bar{H} \leq U(i) + H(i,j) \\ \forall j \in \mathcal{A}, \ \exists i \in \mathcal{A}, \ U(j) + \bar{H} \geq U(i) + H(i,j) \\ \forall j \in \mathcal{A}, \ U(j) = \min\{U(i) + H(i,j) : i \in \mathcal{A}\} \end{cases}$$

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Zero limit : Ground states of a directed graph

Conclusion We just have proved that $\bar{H}_{\beta} \to \bar{H}$ and $U_{\beta} \to U$ $T[U] = U + \bar{H}$ $T[U](j) := \min_{i \in \mathcal{A}, i \to j} (U(i) + H(i, j))$

We extend U as a function on the SFT X

$$u(x) = U(x_0), \quad x = (x_k)_{k \ge 0}$$

We extend H as a function on X

$$\phi(x) = H(x_0, x_1), \quad x = (x_k)_{k \ge 0}$$

Then

$$T[u] = u + \bar{H}$$

$$T[u](y) = \min_{x:f(x)=y} (u(x) + \phi(x))$$

By uniqueness of the additive eigenvalue

$$\bar{H} = \bar{\phi}$$

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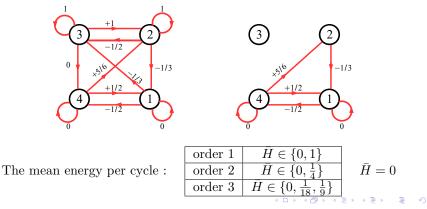
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Zero limit : Ground states of a directed graph

Question Can we compute explicitly \overline{H} ?

Proposition

- (1) \bar{H} equals the minimum of the mean energy over all simple cycles
- (2) the minimizing measures are supported on the SFT made of minizing cycles



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Zero limit : Ground states of a directed graph

Proof

1 We have shown the existence of a calibrated subaction U

$$\begin{cases} \forall \, i \to j \in \mathcal{A}, & U(j) + \bar{H} \leq U(i) + H(i,j) \\ \forall \, i_0 \in \mathcal{A}, \ \exists \, i_{-1} \in \mathcal{A}, & U(i_0) + \bar{H} = U(i_{-1}) + H(i_{-1},i_0) \end{cases}$$

2 we construct a backward orbit that calibrates H

$$\exists i_{-n} \to i_{-(n-1)} \to \cdots i_{-1} \to i_0$$
$$U(i_{-k}) + \bar{H} = U(i_{-k-1}) + H(i_{-k-1}i_{-k})$$

3 because the graph is finite the backward orbit closes up

$$\exists p \ge 1, \ i_{-n-p} = i_{-n}$$

4 by telescoping sum U disappears

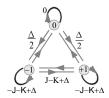
III. Zero temperature limit in thermodynamic formalism

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Zero limit : Explicit computation for BEG

The BEG model



Mean of H along simple cycles :

cycles of order 1	$0, (-J - K + \Delta)$
cycles of order 2	$\frac{1}{2}\Delta, (J-K+\Delta)$
cycles of order 3	$\frac{1}{3}(J-K+2\Delta)$

The energy matrix is

$$M_{\beta} = \begin{bmatrix} \exp\left(-\beta(-J-K+\Delta)\right) & \exp\left(-\beta(\frac{1}{2}\Delta)\right) & \exp\left(-\beta(J-K+\Delta)\right) \\ \exp\left(-\beta(\frac{1}{2}\Delta)\right) & 0 & \exp\left(-\beta(\frac{1}{2}\Delta)\right) \\ \exp\left(-\beta(J-K+\Delta)\right) & \exp\left(-\beta(\frac{1}{2}\Delta)\right) & \exp\left(-\beta(-J-K+\Delta)\right) \end{bmatrix}$$

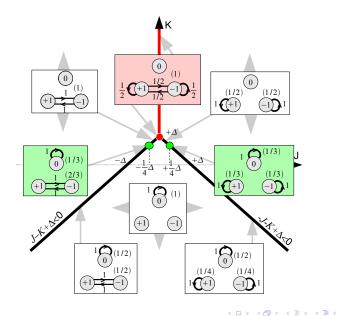
We discuss the phase diagram according to the smallest term

$$\min\left(0,\frac{\Delta}{2},-J-K+\Delta,J-K+\Delta,\frac{1}{3}(J-K+2\Delta)\right)$$

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Zero limit : Explicit computation for BEG



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- I. Introduction
- II. Additive ergodic optimization on hyperbolic spaces
- III. Zero temperature limit in thermodynamic formalism
- IV. Discrete Aubry-Mather and Frenkel-Kontorova model
- V. Contreras genericity of periodic orbits
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IV. Discrete Aubry-Mather and the Frenkel-Kontorova model

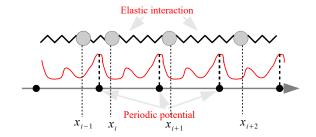
• The Frenkel-Kontorova model

- Calibrated configurations
- The algorithm

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Discrete Aubry-Mather : The Frenkel-Kontorova model

The physical model The model describes the set of configuration of a chain of atoms at equilibrium in a periodic external environment



The original 1D-FK

 $E_{\lambda,K}(x,y) = W_{\lambda}(x,y) + V_{K}(x), \quad x,y \in \mathbb{R}$ $W_{\lambda}(x,y) = \frac{1}{2\tau} |y - x - \lambda|^{2} - \frac{\lambda^{2}}{2\tau}, \quad V_{K}(x) = \frac{K\tau}{(2\pi)^{2}} \Big(1 - \cos(2\pi x) \Big)$ $E_{\lambda,K}(x,y) = E_{0,K}(x,y) - \lambda(y - x)$

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Discrete Aubry-Mather : The Frenkel-Kontorova model

Question Is it possible to define a notion of configurations $\underline{\mathbf{x}} := (x_k)_{k \in \mathbb{Z}}, x_k \in \mathbb{R}$, with the smallest total energy

$$E_{tot}(\underline{\mathbf{x}}) := \sum_{k=-\infty}^{+\infty} E(x_k, x_{k+1}) \le E_{tot}(\underline{\mathbf{y}}), \quad \forall \ \underline{\mathbf{y}} = (y_k)_{y \in \mathbb{Z}}$$

Definition A configuration $(x_n)_{n \in \mathbb{Z}}$ is said to be minimizing if the energy of a finite block of atoms with two fixed extremities cannot be lowered by displacing atoms inside the block :

• define
$$E(x_m, x_{m+1}, \dots, x_n) := \sum_{k=m}^{n-1} E(x_k, x_{k+1})$$

• if $(y_m, y_{m+1}, \dots, y_n)$ is another configuration with the two endpoints fixed, $y_m = x_m$ and $y_n = x_n$ then

$$E(x_m, x_{m+1}, \dots, x_n) \le E(y_m, y_{m+1}, \dots, y_n)$$

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Discrete Aubry-Mather : The Frenkel-Kontorova model

Remark The notion of minimizing configurations is NOT correct. Consider

$$E_{\lambda}(x,y) := E(x,y) - \lambda \cdot (y-x)$$

 $(\lambda$ is the distance between the atoms at rest). Then

 $(x_k)_{k\in\mathbb{Z}}$ is minimizing for $E_\lambda \Leftrightarrow (x_k)_{k\in\mathbb{Z}}$ is minimizing for E_0

Proof

$$\sum_{k=m}^{n-1} \left(E_0(x_k, x_{k+1}) - \lambda(x_{k+1} - x_k) \right) = \sum_{k=m}^{n-1} E_0(x_k, x_{k+1}) - \lambda(x_n - x_m)$$

Remarks

- (1) minimal geodesics have a similar definition (λ is a cohomological factor)
- (2) minimizing configurations look like local minimizers of some functional energy. We need a stronger notion of global minimizers that will be called calibrated configurations

IV. Discrete Aubry-Mather and the Frenkel-Kontorova model

- The Frenkel-Kontorova model
- Calibrated configurations
- The algorithm

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Discrete Aubry-Mather : Calibrated configurations

Definition The effective energy of a configuration is

$$\bar{E} := \lim_{n \to +\infty} \inf_{x_0, \dots, x_n \in \mathbb{R}^d} \frac{1}{n} \sum_{k=0}^{n-1} E(x_k, x_{k+1})$$

Remark

- The limit exists by super-additivity
- By coercitivity of E(x, y): $\lim_{|y-x|\to+\infty} E(x, y) = +\infty$

$$-\infty < \inf_{x,y \in \mathbb{R}} E(x,y) \le \overline{E} \le \inf_{x \in \mathbb{R}^d} E(x,x) < +\infty$$

Definition

• The Mañé potential between two positions $x, y \in \mathbb{R}$ is

$$S(x,y) := \inf_{n \ge 1} \inf_{x = x_0, \dots, x_n = y} \sum_{k=0}^{n-1} \left(E(x_k, x_{k+1}) - \bar{E} \right)$$

• $\underline{x} = (x_k)_{k \in \mathbb{Z}}$ is said to be calibrated if

$$\forall m < n, \quad \sum_{k=m}^{n-1} \left(E(x_k, x_{k+1}) - \bar{E} \right) = S(x_m, x_n)$$

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Discrete Aubry-Mather : Calibrated configurations

Question How can we find calibrated configurations?

The Lax-Oleinik operator For every periodic function $u: \mathbb{R} \to \mathbb{R}$

$$T[u](y) := \inf_{x \in \mathbb{R}} \left(u(x) + E(x, y) \right)$$

Remark

- By coercivity of E, the infimum is attained
- We have chosen an interaction energy satisfying

$$E(x+1,y+1) = E(x,y)$$

• In particular : u periodic $\Rightarrow T[u]$ periodic

Theorem There exists a Lipschitz periodic function $u : \mathbb{R} \to \mathbb{R}$ solution

$$T[u] = u + \bar{E}$$

u is called effective potential. It is not unique. The additive eigenvalue \bar{E} is unique

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Discrete Aubry-Mather : Calibrated configurations

Construction of calibrated configurations

- **1** solve $T[u](y) = u(y) + \overline{E} = \min_x (u(x) + E(x, y))$
- **2** choose $x_0 \in [0, 1]$ and construct a backward optimal configuration

$$u(x_{-k}) + \bar{E} = u(x_{-k-1}) + E(x_{-k-1}, x_{-k})$$

- **3** shift the finite configuration $(x_k + L_n)_{k=-2n}^0$ by an integer L_n so that $x_{-n} + L_n \in [0, 1]$
- ④ extract a convergent subsequence $(x_k^\infty)_{k \in \mathbb{Z}}$ by a diagonal argument
- **5** the limit $(x_k^{\infty})_{k \in \mathbb{Z}}$ is calibrated

Discrete Aubry-Mather : Calibrated configurations

Theorem Recall $E_{\lambda}(x,y) = E_0(x,y) - \lambda(y-x), \quad \underline{x} = (x_k)_{k \in \mathbb{Z}}$

- (1) \underline{x} is minimizing for $E_{\lambda} \Leftrightarrow \underline{x}$ is minimizing for E_0
- (2) A calibrated configuration for E_{λ} is minimizing
- (3) A minimizing configuration is calibrated for some E_{λ}
- (4) Recall

$$\bar{E}(\lambda) := \lim_{n \to +\infty} \inf_{x_0, \dots, x_n \in \mathbb{R}^d} \frac{1}{n} \sum_{k=0}^{n-1} E_{\lambda}(x_k, x_{k+1})$$

- (5) $\lambda \mapsto \overline{E}(\lambda)$ is a C^1 function
- (6) A calibrated configuration for E_{λ} admits a rotation number

$$\lim_{n \to \pm \infty} \frac{x_n - x_0}{n} = \omega(\lambda) := -\frac{dE}{d\lambda}$$

(7) Emergence of the locking phenomena at rational rotation number

$$\operatorname{Leb}\left(\mathbb{R}\setminus\bigcup_{p/q\in\mathbb{Q}}\operatorname{interior}\left\{\lambda\in\mathbb{R}:\omega(\lambda)=\frac{p}{q}\right\}\right)=0$$

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IV. Discrete Aubry-Mather and the Frenkel-Kontorova model

- The Frenkel-Kontorova model
- Calibrated configurations
- The algorithm

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Discrete Aubry-Mather : The algorithm

The 1D-FK model

$$E_{\lambda,K}(x,y) := \frac{1}{2\tau} |y-x|^2 - \lambda(y-x) + \frac{K\tau}{(2\pi)^2} \left(1 - \cos(2\pi x)\right)$$

Ishikawa's algorithm

- **1** discretize the initial cell $[0,1], z_i = \frac{i}{N}, i = 1, \dots, N$
- **2** choose a number of cells around the initial cell $R \ge 1$
- **3** start with the zero potential $u_0 = 0$. Assume u_n is known
- **4** construct the optimal backward map

$$z_j \mapsto (z_{\tau(j)}, p_j) = \operatorname*{arg\,min}_{z_i, \ p \in \llbracket -R, R \rrbracket} \left(u_n(z_i) + E_{\lambda, K}(z_i + p, z_j) \right)$$

6 compute Lax-Oleinik

$$T[u_n](z_j) = u_n(z_{\tau(j)}) + E_{\lambda,K}(z_{\tau(j)} + p_{\tau(j)}, z_j)$$

6 use Ishikawa's algorithm

$$u_{n+1} = \frac{u_n + T[u_n]}{2} - \min\left(\frac{u_n + T[u_n]}{2}\right)$$

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Discrete Aubry-Mather : The algorithm

Ishikawa's algorithm

- \circ stop the algorithm until $\max_i |u_{n+1}(z_i) u_n(z_i)| \le \epsilon$
- 8 compute the backward minimizing cycle

$$i_0 \to i_1 = \tau(i_0), p_1 \to i_2 = \tau(i_1), p_2, \to \cdots$$

9 choose the smallest $q \ge 1$ such that $i_q = i_0$,

$$\textbf{() define } p = p_1 + \dots + p_q$$

- **(1)** the rotation number equals $\omega = \frac{p}{q} = -\frac{1}{\tau} \frac{\partial \bar{E}}{\partial \lambda}$
- (2) the Mather set is the periodic orbit

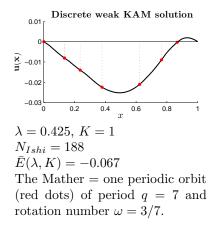
$$z_{i_0}, z_{i_1}, \ldots, z_{i_q}$$

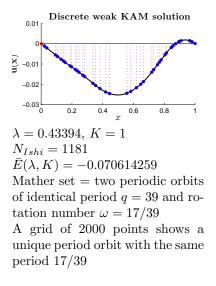
Choice of the constants

•
$$\tau = 1, N = 1000, R = 2, \epsilon = 10^{-9}$$

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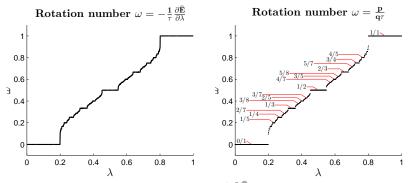
Discrete Aubry-Mather : The algorithm





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Discrete Aubry-Mather : The algorithm

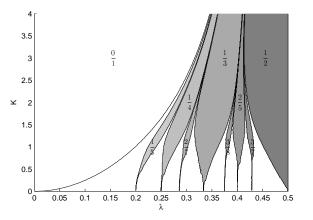


Graph of the rotation number $\omega = -\frac{1}{\tau} \frac{\partial E}{\partial \lambda}(\lambda)$ (lefthand side), and $\omega = \frac{p(\lambda)}{\tau q(\lambda)}$ (right hand side). The coupling is K = 1, the grid on λ is 0: 0.0005: 1. The maximum number of iteration is 198, the maximum jump is 1.286, the maximum number of cycles is 2.

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Discrete Aubry-Mather : The algorithm



Phase diagram of the Frenkel-Kontorova model : $\tau = 1$, N = 400, $\lambda = 0 : 0.001 : 0.5$, K = 0 : 0.01 : 4. Each domain is parametrized by a rotation number $\omega = \frac{p}{\tau q}$

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Bibliographie I

- A. Fathi, The weak KAM theorem in Lagrangian dynamics, book to appear, Cambridge University Press.
- E. Garibaldi, Ph. Thieullen. Minimizing orbits in the discrete Aubry–Mather model. Nonlinearity, Vol. 24 (2011), 563–611.
- E. Garibaldi, Ph. Thieullen. Description of Some Ground States by Puiseux Techniques. J. Stat. Phys. Vol. 146 (2012), 125–180.
- E. Garibaldi. Ergodic Optimization in the Expanding Case Concepts, Tools and Applications. Springer 2017.
- O. Jenkinson. Ergodic optimization in dynamical systems. Ergod. Th. and Danym. Sys. Vol. 39 (2019), 2593–2618.
- R. Jungers. The Joint Spectral Radius. Theory and Application. Springer 2009.