

Ergodic Optimization and Prevalence

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Ergodic optimization: an overview

Basic reference: O. Jenkinson. *Ergodic Optimization*, Discrete and Cont. Dyn. Sys. A, vol. 15 (2006), pp. 197–224.

Disclaimer: I won't discuss the relations with Classical Mechanics or Thermodynamical Formalism.

Ergodic optimization: the general setting

- $X =$ compact metric space
- $T: X \rightarrow X$ continuous map
- $f: X \rightarrow \mathbb{R}$ continuous function (“performance” or “potential”)
- $\mathcal{M}_T := \{T\text{-invariant probability measures}\}$
- “ergodic supremum”

$$\begin{aligned} \beta(f) &:= \sup_{\mu \in \mathcal{M}_T} \int f d\mu \\ &= \sup_{x \in X} \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) \\ &= \lim_{n \rightarrow \infty} \sup_{x \in X} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) \end{aligned}$$

An easy example

$X = \{0, 1\}^{\mathbb{N}} = 2^{\mathbb{N}}$ Cantor set

$T: 2^{\mathbb{N}} \rightarrow 2^{\mathbb{N}}$ shift

$f =$ characteristic function of cylinder $C = [101]$

Then $\beta(f) = 1/2$. Indeed:

- Since $T^{-1}(C) \cap C = \emptyset$, for every $x \in 2^{\mathbb{N}}$, the frequency of visits to C is $\leq 1/2$;
- The T -invariant prob. μ supported on the orbit of $\overline{10} = (1, 0, 1, 0, \dots)$ has $\int f d\mu = 1/2$.
Rem.: μ is the *unique* such measure.

Maximizing measures

In general, a measure $\mu \in \mathcal{M}_T$ s.t. $\int f d\mu = \beta(f)$ is called a **maximizing measure**.

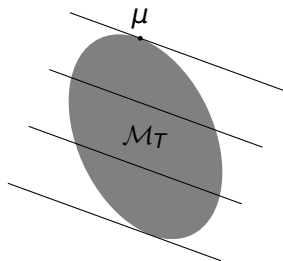
Existence? Yes (compactness).

Generic uniqueness:

Theorem (Jenkinson and others)

For (topologically) generic f in any “reasonable”(*) space \mathcal{F} of continuous functions, the maximizing measure is **unique**.

(*) a vector space \mathcal{F} continuously and densely embedded in $C^0(X)$.



The general problem

Problem

For a fixed “nice” dynamical system T , and a fixed “nice” family/space \mathcal{F} of functions f , understand the maximizing measures for all/most functions f .

Of course, the problem is uninteresting if T has few invariant measures.

In most of the literature, T is assumed to have strong **hyperbolicity** properties and therefore lots of periodic measures.

In all that follows we will assume T to be **uniformly expanding**.

Regularity makes a big difference

Assume $T =$ uniformly expanding.

Theorem (Bousch–Jenkinson)

For generic C^0 functions, the maximizing measures have full support.

The situation is very different if the functions are more regular:

Theorem (Subordination principle)

If $f \in C^\alpha$ (i.e. f is α -Hölder) then there exists a compact invariant set $K_f \subset X$ (“Mather set”) such that

$$\mu \in \mathcal{M}_T \text{ is maximizing for } f \iff \text{supp } \mu \subset K_f.$$

Corollary of the **Mañé Lemma** (or Mañé–Conze–Guivarc’h–Savchenko–Fathi–Contreras–Lopes–Thieullen–Bousch Lemma).

A nice example (Hunt, Ott, Jenkinson, Bousch)

The following example was first studied experimentally by Hunt and Ott (1996):

- $T(x) := 2x \bmod 1$ on $X = \mathbb{R}/2\pi\mathbb{Z}$.
- Family \mathcal{F} of functions: (nonzero) linear combinations of $\cos x$ and $\sin x$.

Theorem (Bousch 2000)

*In that setting, maximizing measures are always unique. Moreover, for an **open and full measure** subset of \mathcal{F} , the maximizing measure is supported on a **periodic orbit**.*

(Actually the maximizing measures are *Sturmian*.)

The big conjecture

Conjecture (Hunt–Ott 1996)

For *typical chaotic* systems, *typical* parameterized families of *smooth* functions, and *most* values of the parameter, the maximizing measure is unique and **supported on a periodic orbit**.

(Terms in color are left undefined. . .)

An important result

Improving on the work of previous authors (Yuan–Hunt, Contreras–Lopes–Thieullen, Bousch, Bressaud–Quas, Morris, Quas–Siefken), Contreras managed to prove the following:

Theorem (Contreras 2013)

For uniformly expanding dynamics, and (topologically) generic Lipschitz functions, the maximizing measure is (unique and) supported on a periodic orbit.

Actually the conclusion holds for an open and dense subset of $C^{\text{Lip}}(X)$, and the “locking property” holds: the maximizing measures are robust under perturbations.

Goal

We would like to obtain results like Contreras', but with genericity being not only in the topological sense, but in a **probabilistic** sense as well (thus being a little closer to the spirit of the Hunt–Ott conjecture).

Setting for our main result (details later):

- T = one-sided shift on 2 symbols;
- \mathcal{F} = space of “super-continuous” functions (very strong modulus of regularity);
- “probabilistic genericity” is expressed in terms of **prevalence**.

Motivation for prevalence

Is it possible to speak of probabilities in infinite-dimensional vector spaces?

- ☹ There is no useful (say, σ -finite) translation-invariant measure.
- ☹ There is no useful (say, σ -finite) translation-invariant class of measures;
- 😊 However there is a translation-invariant notion of “almost every point”, called **prevalence** [Hunt–Sauer–Yorke, Christensen].

Measure transversality and shyness

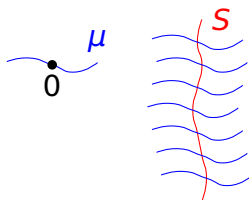
- \mathcal{F} = complete metrizable vector space;
- $S \subset \mathcal{F}$ Borel set;
- μ = Borel probability measure on \mathcal{F} with compact support.

μ is called **transverse** to S ($\mu \bar{\cap} S$) if:

$$\forall f \in \mathcal{F}, \quad \mu(S - f) = 0.$$

I.e. summing to *any* $f \in \mathcal{F}$ a random perturbation we get outside of S with μ -probability 1.

$S \subset \mathcal{F}$ is called **shy** if $\exists \mu \bar{\cap} S$.



Prevalent sets

A Borel subset of a complete metrizable vector space is called **prevalent** if its complement is shy.

Less formally: In order to prove that a set $P \subset \mathcal{F}$ is prevalent, we need to find a compactly supported measure μ such that given *any* $f \in \mathcal{F}$, if we perturb f by adding a μ -random term g , then $f + g \in P$ with μ -probability 1.

In that case, we can always replace μ by another with small support. Thus $f + g$ can be thought as a **random perturbation** of f .

Properties of prevalence

- $\dim \mathcal{F} < \infty \Rightarrow$ the prevalent sets are exactly those of full Lebesgue measure.
- Prevalence is preserved under translation.
- Prevalence is preserved under augmentation.
- Prevalence is preserved under countable intersection.
- Prevalence implies denseness.

Some spaces of functions on $2^{\mathbb{N}}$

Given a sequence of positive numbers $\mathbf{a} = (a_n) \searrow 0$,
define a metric on $2^{\mathbb{N}} = \{0, 1\}^{\mathbb{N}}$:

$$d_{\mathbf{a}}(x, y) := a_{n(x, y)} \quad \text{where} \quad n(x, y) := \inf\{i \in \mathbb{N}; x_i \neq y_i\}.$$

Space of functions:

$$C^{\mathbf{a}}(2^{\mathbb{N}}) := \{f: X \rightarrow \mathbb{R}; f \text{ is Lipschitz w.r.t. } d_{\mathbf{a}}\}$$

(The faster $a_n \rightarrow 0$, the smaller the space $C^{\mathbf{a}}$.)

This is a (nonseparable) Banach space with the norm:

$$\|f\|_{\mathbf{a}} := \|f\|_{\infty} + \text{Lip}_{\mathbf{a}}(f).$$

Example: $d_{\mathbf{a}}$ with $\mathbf{a} = (2^{-n})$ is the “usual” metric on X .
The space of α -Hölder functions w.r.t. the usual metric
is $C^{\mathbf{b}}(2^{\mathbb{N}})$ where $\mathbf{b} = (2^{-\alpha n})$.

The main theorem

Theorem (joint with Yiwei Zhang. ArXiv:1501.00961)

The locking property (*) is prevalent in $C^a(2^{\mathbb{N}})$, provided $\mathbf{a} = (a_n) \searrow 0$ sufficiently fast (**).

(*) A function $f \in C^a(2^{\mathbb{N}})$ satisfies the **locking property** if:

- f has a unique maximizing measure μ (w.r.t. the shift), and it is periodic;
- μ is also the unique maximizing measure for every $g \in C^a(2^{\mathbb{N}})$ sufficiently close to f .

(**) Unfortunately, we need really fast convergence to 0, namely:

$$\frac{a_{n+1}}{a_n} = o\left(2^{-2^{n+2}}\right)$$

Haar series

Every continuous function f on the Cantor set $2^{\mathbb{N}}$ has a uniformly convergent (*) **Haar series**:

$$f(x) = c + \sum_{\omega} c_{\omega} h_{\omega}(x),$$

where ω runs on the (finite) words on the letters 0, 1.

(*) In that sense Haar series are better behaved than Fourier series.

The spaces $C^a(2^{\mathbb{N}})$ introduced before can be essentially characterized in terms of the decay of the Haar coefficients (c_{ω}) .

The random perturbations

Given a family of positive numbers $\mathbf{b} = (b_\omega)$ indexed by words ω , we define a set of functions:

$$\mathcal{H}_{\mathbf{b}} := \left\{ \sum_{\omega} c_{\omega} h_{\omega}; c_{\omega} \in [-b_{\omega}, b_{\omega}] \right\} = \mathbf{Hilbert\ brick.}$$

Then, for appropriate \mathbf{b} (e.g. $b_{\omega} = a_n/(n+1)$, $n = |\omega|$):

- $\mathcal{H}_{\mathbf{b}}$ is a compact subset of $C^a(2^{\mathbb{N}})$;
- taking random independent coefficients $c_{\omega} \sim \text{Uniform}([-b_{\omega}, b_{\omega}])$ we obtain a probability $\mu_{\mathbf{b}}$ supported on $\mathcal{H}_{\mathbf{b}}$;
- **these are the random perturbations in our Main Theorem**, i.e., the measure $\mu_{\mathbf{b}}$ is transverse to the set of functions that don't have the locking property.

Strategy of the proof of the Main Theorem

A **step function** of level n is a function on $2^{\mathbb{N}}$ that is constant on cylinders of rank n . We will see that **step functions have periodic maximizing measures**.

Since $\mathbf{a} = (a_n) \rightarrow 0$ very fast, the functions f in $C^{\mathbf{a}}(2^{\mathbb{N}})$ are well-approximated by step functions f_n (which can be obtained by truncating the Haar series).

We will show that **with probability 1** (in any translated Hilbert brick. . .), **the maximizing measure for f coincides with the (periodic) maximizing measure for f_n** for some n .

We need quantitative information on the ergodic optimization of step functions. . .

Finite dimensional ergodic optimization

Let F be a finite-dimensional vector space of functions, with basis $\{f_1, \dots, f_n\}$.

Define a “projection” linear map $\pi: \mathcal{M} \rightarrow \mathbb{R}^n$ on the vector space of signed measures \mathcal{M} by

$$\pi(\mu) := \left(\int f_1 d\mu, \dots, \int f_n d\mu \right).$$

Define a compact convex set:

$$R := \pi(\mathcal{M}_T) = \mathbf{rotation\ set}$$

(the projection of the T -invariant probability measures).

Origin of the name: $(f_1, \dots, f_n) =$ displacement function of a map $T: \mathbb{T}^n \rightarrow \mathbb{T}^n$ homotopic to id.

Finite dimensional ergodic optimization

Functions $f \in F$ can be “integrated” with respect to vectors $v \in R = \pi(\mathcal{M}_T)$:

$$\langle f, v \rangle := \int f d\mu \quad \text{where } \mu \text{ is s.t. } \pi(\mu) = v.$$

To compute the “ergodic supremum” becomes a finite-dimensional problem:

$$\beta(f) := \sup_{\mu \in \mathcal{M}_T} \int f d\mu = \sup_{v \in R} \langle f, v \rangle.$$

If the extreme points of the rotation set R happen to have unique preimages in \mathcal{M}_T then every $f \in F$ has a unique maximizing measure.

Finite dimensional ergodic optimization

Conclusion

Ergodic optimization of functions in an n -dimensional space $F \in C^0(X)$ is basically equivalent to:

- regarding F as $(\mathbb{R}^n)^*$;
- determining the extreme points of the compact convex set $R := \pi(\mathcal{M}_T) \subset \mathbb{R}^n$;
- determining their preimages under $\pi: \mathcal{M}_T \rightarrow \mathbb{R}^n$.

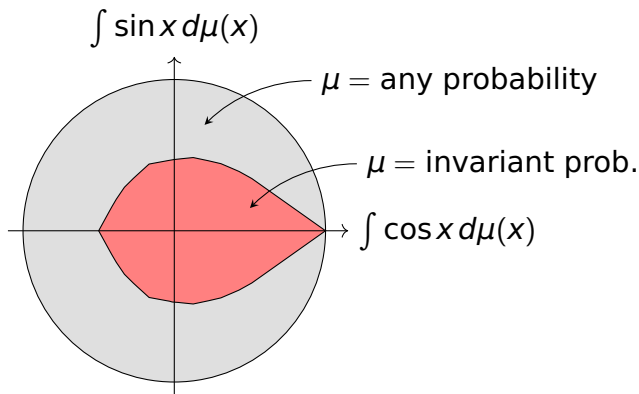
Remark

For $T = \text{shift}$, every compact convex set $R \subset \mathbb{R}^n$ can be realized as a rotation set (for suitable C^0 functions).
(Kucherenko–Wolf)

The fish on the dish

Example #1 (Hunt, Ott, Jenkinson, Bousch):

$T(x) = 2x \bmod 1$ on $\mathbb{R}/2\pi\mathbb{Z}$, $F := \{\text{trig. poly. deg } 1\}$.



Note: “sharper” extreme points of the fish are more likely to be maximizing. . .

Example #2: step functions of level 2

$F := \{\text{step functions on } 2^{\mathbb{N}} \text{ of level 2}\}$,
with basis $\chi_{[00]}, \chi_{[01]}, \chi_{[10]}, \chi_{[11]}$.

The projection $\pi: \mathcal{M} \rightarrow \mathbb{R}^4$ is:

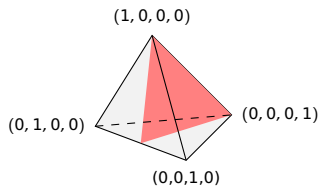
$$\mu \mapsto (\mu([00]), \mu([01]), \mu([10]), \mu([11])).$$

The “dish” $\pi(\{\text{prob. measures}\}) =$
unit simplex:

$$\Delta = \{(p_{ij}) \in \mathbb{R}^4; p_{ij} \geq 0, \sum p_{ij} = 1\}.$$

The “fish” $R = \pi(\{\text{inv. prob.}\})$ is

$$R = \{(p_{ij}) \in \Delta; p_{01} = p_{10}\}.$$



The vertices have
unique pre-images
in $\mathcal{M}_{\mathcal{T}}$, which are
measures supported
on periodic orbits:

Vertex of R	per. orb.
$(1, 0, 0, 0)$	$\overline{0}$
$(0, 0, 0, 1)$	$\overline{1}$
$(0, \frac{1}{2}, \frac{1}{2}, 0)$	$\overline{01}$

Generalization: Step functions of level n

For the shift $T: 2^{\mathbb{N}} \rightarrow 2^{\mathbb{N}}$, consider:

- $F_n := \{\text{step functions of level } n\} \simeq \mathbb{R}^{2^n}$;
- $R_n :=$ associated rotation set.

Theorem (Ziemian)

- The rotation set R_n is a **polytope** in \mathbb{R}^{2^n} ;
- each vertex of R_n is the projection of a **unique** shift-invariant measure, which is supported on a periodic orbit.

The polytopes R_n

	dim	# vertices	assoc. periodic orbits
R_1	1	2	$\overline{0}, \overline{1}$
R_2	2	3	$\overline{0}, \overline{1}, \overline{01}$
R_3	4	6	$\overline{0}, \overline{1}, \overline{01}, \overline{001}, \overline{011}, \overline{0011}$
R_4	8	19	
R_5	16	179	
R_6	32	30166	

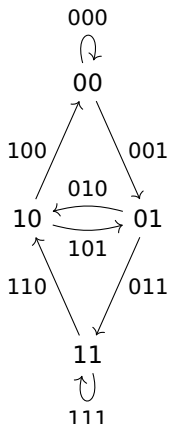
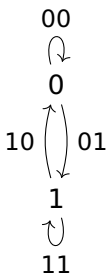
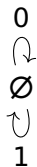
The number of vertices grows super-exponentially; there is no exact formula.

To describe the polytopes R_n , we need to introduce a combinatorial object.

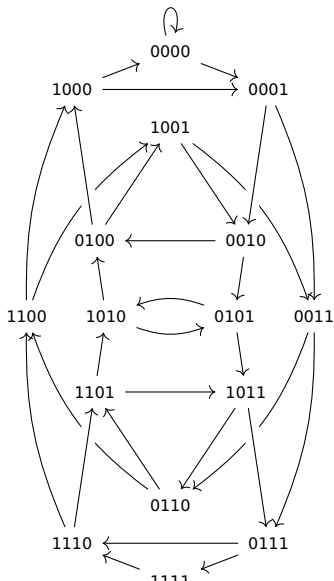
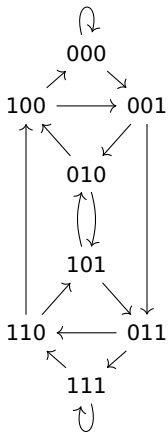
de Bruijn graphs

The **de Bruijn graph** G_n has:

- nodes labelled by words on length $n-1$;
- arrows labelled by words ω on length n , of form $\text{prefix}(\omega) \xrightarrow{\omega} \text{suffix}(\omega)$;



G_4 and G_5



The graph G_n and the rotation set R_n

Recall: $F_n := \{ \text{step functions of level } n \}$

Given $f \in F_n$ assigns **weights** of the arrows of G_n . The maximizing measure μ for f can be obtained as follows:

- find the (simple closed) **cycle of G_n of maximum mean weight**;¹
- this cycle can be seen as a periodic orbit for the shift;
- μ is the measure supported on this orbit.

Conclusion

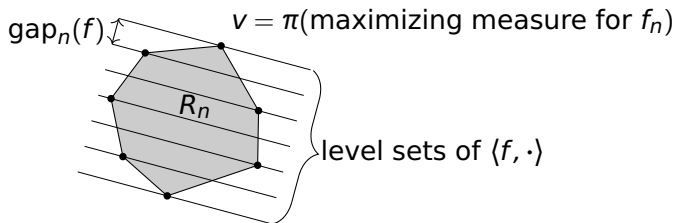
The set R_n is indeed a polytope; its vertices correspond to the (simple closed) cycles on the graph G_n .

¹This problem is studied in applied math (Karp algorithm ...)

A “measure” of uniqueness

Suppose $f: 2^{\mathbb{N}} \rightarrow \mathbb{R}$ is a step function of level n , or equivalently, an attribution of weights to the arrows of G_n .

Compute $\langle f, \nu \rangle$ for each vertex ν of the polytope R_n .
Let $\text{gap}_n(f) :=$ the difference between the maximum and the second maximum:



So $\text{gap}_n(f) \geq 0$, and $\text{gap}_n(f) > 0$ iff the maximizing measure is unique.

Proof of the prevalence theorem

Let us recall the main theorem:

Theorem (B., Zhang)

Fix a space of “super-continuous” functions $C^a(2^{\mathbb{N}})$, and an appropriate Hilbert brick

$$\mathcal{H}_{\mathbf{b}} := \left\{ \sum_{\omega} c_{\omega} h_{\omega}; c_{\omega} \in [-b_{\omega}, b_{\omega}] \right\}.$$

Let $g \in C^a(2^{\mathbb{N}})$, and take a random function f in the translated Hilbert brick $g + \mathcal{H}_{\mathbf{b}}$.

Then there exists a “periodic measure” μ which is the unique maximizing measure for f and for all $\tilde{f} \in C^a(2^{\mathbb{N}})$ sufficiently close to f .

Main Lemma: the Gap criterion

Lemma (Gap criterion)

Given an **arbitrary continuous function** f , truncate its Haar series to obtain a step function f_n :

$$f(x) = c(f) + \sum_{\omega} c_{\omega}(f) h_{\omega}(x) \Rightarrow f_n(x) := c(f) + \sum_{|\omega| < n} c_{\omega}(f) h_{\omega}(x).$$

If the following **gap condition** holds:

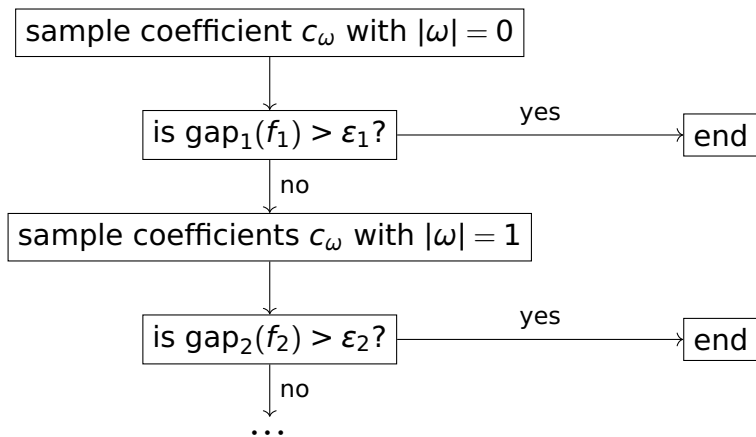
$$\text{gap}_n(f_n) > \sum_{k=n}^{\infty} (k - n + 1) \max_{|\omega|=k} |c_{\omega}(f)|$$

then the maximizing measure for f_n (which is unique and periodic) is also the maximizing measure for f .

Proof: Combinatorial arguments with the de Bruijn graphs (4 pages).

Proof of the theorem

Let ε_n be an upper bound for the RHS in the gap condition. The following “algorithm” finds the maximizing measure (provided it stops):



Proof of the theorem

We need to show that the algorithm stops with probability 1, i.e., $\text{Prob}[\exists n; \text{gap}_n(f_n) > \varepsilon_n] = 1$.

- $\text{gap}_n(f_n)$ depends on the Haar coefficients of level $n - 1$;
- $\varepsilon_n = O(\text{Haar coefficients of level } n)$;
- the Haar coefficients of level n are much smaller than the **variance** of the Haar coefficients of level $n - 1$.

It follows that:

- $\text{variance}(\text{gap}_n(f_n)) \gg \varepsilon_n$;
- $\text{Prob}[\text{gap}_n(f_n) > \varepsilon_n] \rightarrow 1$ (overkill)
- $\text{Prob}[\text{algorithm stops at a level } \leq n] \rightarrow 1$

Proof of the theorem

Why do we need super-exponential decay of the Haar coefficients (strong modulus of continuity)?

Because:

- the polytope R_n has a super-exponential number of vertices;
- these vertices are the candidates for maximizing measures for f_n ;
- and we need to guarantee a gap between the top 2 vertices.

How to improve the main result?

What about the **Hölder case** (exponential decay of Haar coefficients)? Recall:

Lemma (Gap criterion)

$$\text{gap}_n(f_n) > \varepsilon_n \geq \sum_{k=n}^{\infty} (k - n + 1) \max_{|\omega|=k} |c_\omega(f)| \quad \Rightarrow$$

the maximizing measure for f_n (which is unique and periodic) is also the maximizing measure for f .

Hölder case $\Rightarrow \varepsilon_n \rightarrow 0$ exponentially, while computer experiments indicate that $\text{gap}_n(f_n) \rightarrow 0$ polynomially (i.e. $O(1/n^\alpha)$) a.s. (despite the super-exponential number of candidate maximizers.)

A finer understanding of the geometry of the polyhedra R_n may help...