

Lecture 2: Symbolic Dynamics of MME

Last Time we stated without proof that BHC can carry at most one MME.

Today we will sketch the proof:

- (a) We'll describe the symbolic dynamics of f on a BHC
- (b) We'll calculate the form of the MME in symbolic coordinates
- (c) We'll deduce uniqueness (and much more).

Markov Shifts

Setup: Suppose we're given

- a countable set of "states" S
- a collection of directed edges $E \subseteq S \times S$ denoted by $a \rightarrow b$ ($(a, b) \in E$).
- $\mathcal{G} :=$ directed graph with vertices S , edges E .

Non-Degeneracy Assumption: We assume throughout that $\forall a \in S \exists \bar{a}, \eta \in S$ s.t. $\bar{a} \rightarrow a, a \rightarrow \eta$.

Markov Shift: $\Sigma(\mathcal{G}) := \{ (x_i)_{i \in \mathbb{Z}} \in S^{\mathbb{Z}} : \forall i \ x_i \rightarrow x_{i+1} \}$
with the metric $d(\underline{x}, \underline{y}) := \exp \left[-\min \{ |i| : x_i \neq y_i \} \right]$
and the action of the left-shift map $\sigma: \Sigma(\mathcal{G}) \rightarrow \Sigma(\mathcal{G})$,
 $\sigma[(x_i)_{i \in \mathbb{Z}}] = (x_{i+1})_{i \in \mathbb{Z}}$.

Exercise: The topology of $\Sigma(\mathcal{G})$ is generated by the base of (non-empty) cylinder sets

$$_m[a_0, \dots, a_n] := \{ \underline{x} \in \Sigma(\mathcal{G}) : x_{i+n} = a_i \ (i=0, \dots, n) \}$$

Exercise:

- (1) $\Sigma(\mathcal{G})$ is compact iff $|S| < \infty$
- (2) $\Sigma(\mathcal{G})$ is locally compact iff \mathcal{G} is **locally finite**:
- $$\forall a \in S, \left(\begin{array}{l} \deg_{in}(a) := \# \{ \tau : \tau \rightarrow a \} < \infty \\ \deg_{out}(a) := \# \{ \tau : a \rightarrow \tau \} < \infty \end{array} \right).$$

Exercise:

- (1) The left shift map is a homeomorphism.
- (2) It's topologically transitive iff $\forall a, b \in S \exists$ non-empty cylinder ${}_0[a, \tau_1, \dots, \tau_{\ell-1}, b]$, $\ell \geq 1$.
In this case we write $a \xrightarrow{\ell} b$
- (3)* It's topologically mixing if it's top transitive, and for some $a \in S$ $\gcd \{ \ell : a \xrightarrow{\ell} a \} = 1$.

Exercise: Let $\Sigma^\#(\mathcal{G}) := \left\{ (x_i)_{i \in \mathbb{Z}} : \begin{array}{l} (x_i)_{i \geq 0}, (x_i)_{i \leq 0} \\ \text{contain constant} \\ \text{subsequences} \end{array} \right\}$.

Prove that $\mu[\Sigma^\#(\mathcal{G})] = 1$ for all inv probability measures μ .

Symbolic Dynamics of Diffeos

The Coding Theorem Suppose $f \in \text{Diff}^{HE}(M)$, $\dim M \geq 2$, M compact. Let H be a BHC. For any $\chi > 0$, there's a Markov shift $\Sigma_\chi = \Sigma(\mathcal{G})$ and a Hölder map $\pi: \Sigma \rightarrow H$ s.t.

$$\begin{array}{ccc} \Sigma_\chi & \xrightarrow{\sigma} & \Sigma_\chi \\ \pi \downarrow & & \downarrow \pi \\ H & \xrightarrow{f} & H \end{array} \quad \text{commutes, and}$$

- (1) Σ_χ is topologically transitive. Moreover, \mathcal{G} is locally finite, and every cylinder B compact.
- (2) $\pi: \Sigma_\chi^\# \rightarrow H$ is finite-to-one
- (3) $\pi(\Sigma_\chi^\#) = H \pmod{\mu}$ for each
 - ergodic inv χ -hyperbolic measure on H
 - projection of any σ -inv measure on Σ_χ
- (4) Hyperbolicity: For some $\chi_0 > 0$, $\forall \underline{x} \in \Sigma_\chi$ there's a splitting $T_{\pi(\underline{x})} M = E^s(\underline{x}) \oplus E^u(\underline{x})$ s.t.
 - $\limsup_{n \rightarrow \infty} \frac{1}{n} \log \|Df^n|_{E^s(\underline{x})}\| < -\chi_0$
 - $\limsup_{n \rightarrow \infty} \frac{1}{n} \log \|Df^{-n}|_{E^u(\underline{x})}\| < -\chi_0$
 - $\underline{x} \mapsto E^s(\underline{x}), \underline{x} \mapsto E^u(\underline{x})$ are Hölder continuous
- (5) Barnological Property: $\exists \varepsilon > 0 \forall (\chi, \varepsilon)$ -Pesin block $\Lambda \subseteq H$, \exists cylinders $A_1, \dots, A_N \subseteq \Sigma$ s.t. $\pi^{-1}(\Lambda) \subseteq \bigcup_{i=1}^N A_i \pmod{\mu}$ for any σ -inv measure μ on Σ_χ .

We omit the proof.

We explore some of the properties of the coding.

Projection Lemma: Any ergodic σ -invariant measure $\hat{\mu}$ on Σ projects to an f -invariant measure on M given by

$$\mu = \pi_* (\hat{\mu}) := \hat{\mu} \circ \pi^{-1}$$

- (1) If $\hat{\mu}$ is ergodic, μ is ergodic
- (2) μ is χ_0 -hyperbolic
- (3) $h_\mu(f) = h_{\hat{\mu}}(\sigma)$

Sketch of Proof: Part (1) is trivial. Part (2) is by the "hyperbolicity property" of π . Part (3) is a consequence of finiteness-to-one, and the following general fact:

Abramov Thm: countable-to-one factors preserve entropy. \square

Lifting Lemma: Any ergodic f -inv χ -hyperbolic measure μ on M is the projection of some σ -inv ergodic measure $\hat{\mu}$ on Σ_χ , with the same entropy.

Wrong Proof: $\hat{\mu} := \mu \circ \pi$. This doesn't work because π is not injective, so $\mu \circ \pi$ is not σ -additive.

Correct Proof (Sketch): Take a.e. ergodic component of

$$\tilde{\mu} := \int_H \frac{1}{|\Sigma^\# \cap \pi^{-1}(y)|} \left(\sum_{\substack{\pi(x)=y \\ x \in \Sigma^\#}} \delta_x \right) d\mu(y)$$

using finiteness-to-one to guarantee the existence of the integral. (Measurability issues are non-trivial.) \square

Corollary: Any χ -hyperbolic MME μ_{\max} lifts to the MME $\hat{\mu}_{\max}$ of the symbolic model $\sigma: \Sigma_{\chi} \rightarrow \Sigma_{\chi}$ of its BHC $H(\mu_{\max})$.

Proof: Firstly, $h_{\mu_{\max}}(\sigma) = h_{\mu_{\max}}(f) = h_{\text{top}}(f)$.

If there were another σ -inv $\hat{\mu}$ s.t. $h_{\hat{\mu}}(\sigma) > h_{\text{top}}(f)$, then its projection $\mu = \pi_{\chi}(\hat{\mu})$ would be an f -inv prob measure with entropy $> h_{\text{top}}(f)$, which contradicts the variational principle.

$\Rightarrow \hat{\mu}_{\max}$ is an MME of $\sigma: \Sigma_{\chi} \rightarrow \Sigma_{\chi}$. \square

Next, we'll show that transitive Markov shifts can have at most one MME, and identify it. This will

- Imply that any BHC can carry at most one MME μ_{\max} and will complete the proof that in dim 2, the number of ergodic MME is finite
- Give us an explicit formula for $\hat{\mu}_{\max}$, opening the way for the proof that $\hat{\mu}_{\max}$, whence μ_{\max} , is Bernoulli up to a period.

Uniqueness and Structure of MME for Transitive Markov Shift

Reduction to One-Sided Shifts: Let \mathcal{G} be a countable directed graph with set of vertices ("states") S . The one-sided Markov shift associated to \mathcal{G} is

$$\Sigma^+ = \Sigma^+(\mathcal{G}) = \left\{ (x_i)_{i \geq 0} : x_i \in S, \forall i \ x_i \rightarrow x_{i+1} \right\}$$
$$\sigma: \Sigma^+ \rightarrow \Sigma^+, \sigma: (x_i)_{i \geq 0} \mapsto (x_i)_{i \geq 1}, d(\underline{x}, \underline{y}) = e^{-\min\{i \geq 0 : x_i \neq y_i\}}$$

Exercise: $\sigma: \Sigma^+(\mathcal{G}) \rightarrow \Sigma^+(\mathcal{G})$ is top. transitive (resp top. mixing) iff $\sigma: \Sigma(\mathcal{G}) \rightarrow \Sigma(\mathcal{G})$ is top. transitive (resp top. mixing).

Let $p: \Sigma \rightarrow \Sigma^+$ be the projection $p: (x_i)_{i \in \mathbb{Z}} \mapsto (x_i)_{i \geq 0}$. Any σ -inv prob measure μ on Σ projects to a σ -inv measure $\mu^+ := \mu \circ p^{-1}$ on Σ^+ given by

$$\mu^+([a_0, \dots, a_{n-1}]) = \mu([a_0, \dots, a_{n-1}]).$$

Conversely, any σ -inv prob. measure μ^+ on Σ^+ extends uniquely to a σ -inv prob. measure μ on Σ called the natural extension of μ^+ and given by

$$\mu([a_0, \dots, a_n]) = \mu^+([a_0, \dots, a_{n-1}])$$

Note that $\mu \circ p^{-1} = \mu^+$.

Exercise: $h_\mu(\sigma) = h_{\mu^+}(\sigma)$

Corollary: μ is MME $\Leftrightarrow \mu^+$ is MME. Thus, It's enough to study MMEs of one-sided Markov shifts.

Ruelle's Operator: Fix some $\phi: \Sigma^+(\gamma) \rightarrow \mathbb{R}$ Hölder cts

(e.g. $\phi \equiv 0$). Ruelle's Operator is

$$(L_\phi f)(x) = \sum_{\sigma y = x} e^{\phi(y)} f(y) \equiv \sum_{a: a \rightarrow x_0} e^{\phi(a, x_0, x_1, \dots)} f(a, x_0, x_1, \dots)$$

(the domain depends on the context).

We will see that L_ϕ is intimately tied to the variational problem of maximizing $h_m(\sigma) + \int \phi dm$.

Let $P_\phi(\phi) := \sup \{ h_m(\sigma) + \int \phi dm \}$ (the "Gurevich pressure" of ϕ).

Variational Principle: Let Σ^+ be a top transitive Markov shift with finite top entropy, and let $\phi: \Sigma^+ \rightarrow \mathbb{R}$ be a (bounded) Hölder function. Then for any $a \in S$, $x_a \in [a]$,

$$P_\phi(\phi) = \overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \log (L_\phi^n 1_{[a]})(x_a)$$

Exercise (Gurevich Variational Principle): Let Σ^+ be a top transitive Markov shift such that $\sup_m h_m(\sigma) < \infty$. * Then

$$\forall a, b \in S, \sup \{ h_m(\sigma) : \begin{array}{l} m \text{ inv Borel} \\ \text{prob. meas.} \end{array} \} = \overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \log t_{a,b}^{(n)},$$

where $t_{ab}^{(n)}$ are the entries of A^n , and A is the transition matrix of Σ^+ , i.e. $A = (t_{ab})_{S \times S}$, $t_{a,b} = \begin{cases} 1 & a \rightarrow b \\ 0 & \text{else} \end{cases}$.

* Exercise*: Remove the condition on the finiteness of $\sup_m h_m(\sigma)$.

Generalized "Ruelle's Perron-Frobenius Thm": Let $\lambda = \exp P_\phi(\phi)$.

Under the prev. assump., if $\exists a \in S, \exists x_a \in [a]$ s.t. $\sum_{n=1}^{\infty} \lambda^{-n} (L_\phi^n 1_{[a]})(x_a) = \infty$ ("ϕ is recurrent"), then \exists Borel measure ν and \exists positive continuous function h s.t. $L_\phi^* \nu = \lambda \nu$ and $L_\phi h = \lambda h$. Moreover,

- (1) ν is finite and positive on cylinders, and without wandering sets of positive measure (but perhaps $\nu(\Sigma^+) = \infty$)
- (2) $\log h, (\log h) \circ \sigma$ are uniformly Hölder on cylinders. But perhaps $\int h d\nu = \infty$.
- (3) If $\phi = \text{const}$, h is constant on cylinders (i.e. $h(x) = h(x_0)$).

Structure Thm for Equilibrium Measures: Suppose

- (Σ^+, σ) is a top. transitive one-sided Markov shift with finite top entropy ($\sup \{h_\mu(\sigma)\} < \infty$).
- $\phi: \Sigma^+ \rightarrow \mathbb{R}$ is a bounded Hölder continuous function
- m_ϕ is an equilibrium measure:

$$h_{m_\phi}(\sigma) + \int \phi d m_\phi = \sup \{ h_\mu(\sigma) + \int \phi d \mu : \mu \text{ } \sigma\text{-inv} \}$$

Then:

- (1) Structure: $m_\phi = h \cdot d\nu$ where $h \geq 0$ is cts, ν is a Borel measure finite and positive on cylinders, and

$$L_\phi h = \lambda h, L_\phi^* \nu = \lambda \nu, \int h d\nu = 1, \lambda = \exp(P_\phi(\phi))$$

- (2) Ergodic Properties: The equilibrium is ergodic and its natural extension is Bernoulli up to a period.

In the special case of MMEs ($\phi \equiv 0$), it's Markov.

- (3) Uniqueness: There's at most one equilibrium measure.

 MME.

The proof requires some preparation.

Sub-Eigenfunctions: For any Hölder continuous $\phi \exists h$ positive continuous s.t. $L_\phi h \leq \lambda h$, $\lambda = \exp P_\phi(\phi)$.

Sketch of Proof:

(1) If ϕ is "recurrent" ($\sum_{n=1}^{\infty} \lambda^{-n} (L_\phi^n 1_{[a]}) (x_a) < \infty$ for some $a \in S$, $x_a \in [a]$), then we can apply the "generalized Birkhoff's Perron-Frobenius theorem" above.

(2) Otherwise $\exists a \in S$, $\exists x_a \in [a]$ s.t. $\sum_{n=1}^{\infty} \lambda^{-n} (L_\phi^n 1_{[a]}) (x_a) < \infty$.
By Hölder continuity $\forall x \in [a]$, $\sum_{n=1}^{\infty} \lambda^{-n} (L_\phi^n 1_{[a]}) (x) < \infty$.

By top transitivity $\forall x \in \Sigma^+$, $\sum_{n=1}^{\infty} \lambda^{-n} (L_\phi^n 1_{[a]}) (x) < \infty$.

Moreover the partial sums are uniformly bounded and equicontinuous on partition sets.

$\Rightarrow h := \sum_{n=1}^{\infty} \lambda^{-n} L_\phi^n 1_{[a]}$ is positive and continuous.

One checks that $L_\phi h \leq \lambda h$. □

The Measure $m \circ \sigma$: Let m be a σ -inv prob. measure on Σ^+ .

Let $\alpha = \{[a] : a \in S\}$ denote the natural partition. Define a Borel measure $m \circ \sigma$ via

$$(m \circ \sigma)(E) := \sum_{A \in \alpha} m(\sigma(E \cap A))$$

Explanation: $\forall A \in \alpha$, $\sigma|_A : A \rightarrow \sigma(A)$ is injective.

But $\sigma : \Sigma^+ \rightarrow \Sigma^+$ is not. The measure $m \circ \sigma$ measures the size of $\sigma(E)$ "with multiplicity:"

Example: If $\Sigma^+ = \{0,1\}^{\mathbb{N}}$ then $\sigma([1]) = \Sigma^+$
but $(m \circ \sigma)([1]) = m(\sigma([1])) + m(\sigma([2])) = 2 \cdot m(\Sigma^+) = 2.$

Exercise:

(1) We cannot define a measure by $\tilde{m}(E) = m(\sigma(E))$,
because (sometimes) $E \mapsto m(\sigma(E))$ is not σ -additive.

(2) If $f : \Sigma \rightarrow \mathbb{R}^+$ is bounded and measurable, then

$$\int_{\Sigma^+} f \, d(m \circ \sigma) = \sum_{a \in S} \int_{\sigma[a]} f(ax) \, dm(x) \quad (ax := (a, x_0, x_1, \dots))$$

(3) If $m \circ \sigma^{-1} = m$, then $m \circ \sigma \ll m$

Solⁿ of Exercise:

(1) If $\Sigma^+ = \{0,1\}^{\mathbb{N}}$, then $\sigma[0] = \sigma[1] = \Sigma^+$ so

$$\tilde{m}([0]) + \tilde{m}([1]) = 2 \cdot m(\Sigma^+) = 2,$$

$$\text{But } \tilde{m}([0] \cup [1]) = \tilde{m}(\Sigma^+) = m(\sigma(\Sigma^+)) = m(\Sigma^+) = 1.$$

(2) It's enough to check the identity for indicators of Borel sets:

$$\int 1_E dm \circ \sigma = (m \circ \sigma)(E) = \sum_{a \in S} m(\sigma(E \cap [a]))$$

$$= \sum_{a \in S} \int 1_{\sigma(E \cap [a])}(y) dm(y)$$

$$\stackrel{!}{=} \sum_{a \in S} \int_{\sigma[a]} 1_E(ay) dm(y).$$

$$y \in \sigma(E \cap [a])$$

$$\Leftrightarrow \sigma|_{[a]}^{-1}(y) \in E \cap [a]$$

$$\Leftrightarrow ay \in E \cap [a]$$

$$\Leftrightarrow y \in \sigma[a], ay \in E$$

(3) We are asked to show that for every Borel set E ,
 $(m \circ \sigma)(E) = 0 \Rightarrow m(E) = 0$. Indeed,

$$(m \circ \sigma)(E) = 0 \Rightarrow \sum_{a \in S} m(\sigma(E \cap [a])) = 0$$

$$\Rightarrow \forall a \in S \quad m(\sigma(E \cap [a])) = 0$$

$$\Rightarrow \forall a \in S \quad (m \circ \sigma^{-1})(\sigma(E \cap [a])) = 0$$

$$m \circ \sigma^{-1} = m \rightarrow$$

$$\Rightarrow \forall a \in S \quad m(E \cap [a]) \leq (m \circ \sigma^{-1})(\sigma(E \cap [a])) = 0$$

$$\Rightarrow m(E) = \sum_{a \in S} m(E \cap [a]) = 0.$$

□

The Jacobian: The Jacobian of an invariant measure μ is
The Radon-Nikodym derivative

$$g_\mu = \frac{d\mu}{d\mu_0} \quad (\mu_0 - \text{a.e.})$$

(I.e. the function g_μ s.t. $\int f g_\mu d\mu_0 = \int f d\mu$.)

Explanation: The "ordinary" Jacobian is the function which appears in the change of variables formula.

The Jacobian g_μ is the "Jacobian" of the "substitution"
 $x = ay = (a, y_0, y_1, \dots)$ for $x \in (a)$:

Fact:

$$\int_{(a)} f d\mu = \int_{\sigma(a)} f(ax) \left(\frac{d\mu}{d\mu_0} \right)(ax) d\mu_0(x)$$

Proof:
$$\int_{(a)} f d\mu = \sum_{\sigma^+} \int_{(a)} \left(1_{(a)} \cdot f \cdot \frac{d\mu}{d\mu_0} \right) \cdot d\mu_0$$

$$= \sum_{b \in S} \int_{\sigma(b)} \left(1_{(a)} \cdot f \cdot \frac{d\mu}{d\mu_0} \right)(bx) d\mu_0(x) \quad (\text{prev. exercise})$$

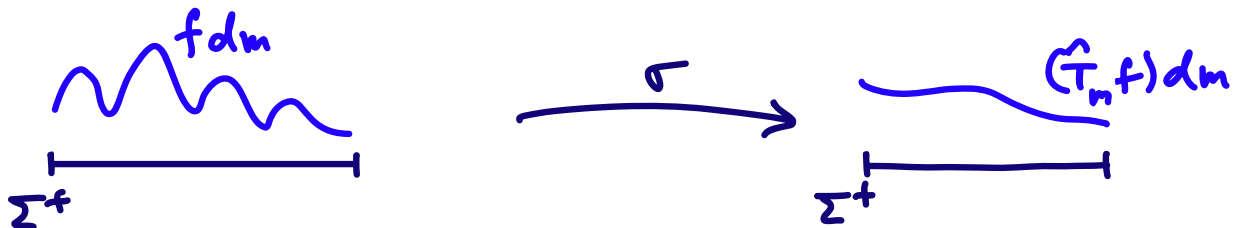
$$= \int_{\sigma(a)} \left(1_{(a)} \cdot f \cdot \frac{d\mu}{d\mu_0} \right)(ax) d\mu_0(x) \quad (\text{all other terms vanish})$$

$$= \int_{\sigma(a)} f(ax) \left(\frac{d\mu}{d\mu_0} \right)(ax) d\mu_0(x).$$

The Transfer Operator: Suppose μ is a σ -invariant prob measure.

The transfer operator of μ is $\hat{T}_\mu: L^1(\mu) \rightarrow L^1(\mu)$ s.t.

$$\sigma_*(f d\mu) = (\hat{T}_\mu f) d\mu$$



Lemma: $(\hat{T}_\mu f)(x) = \sum_{\sigma y = x} g_\mu(y) f(y)$, $g_\mu = \frac{d\mu}{d\mu \circ \sigma}$

Proof: For each $\varphi \in L^1(\mu)$,

$$\int \varphi \cdot \hat{T}_\mu f d\mu \equiv \int \varphi \sigma_*(f d\mu) \equiv \int (\varphi \circ \sigma) \cdot f d\mu$$

$$= \sum_{a \in S} \int \varphi(\sigma x) f(x) d\mu$$

$$= \sum_{a \in S} \int \varphi(\sigma(ax)) f(ax) g_\mu(ax) d\mu(x)$$

$$\left(\text{because } \int \varphi d\mu = \int \varphi(ax) \left(\frac{d\mu}{d\mu \circ \sigma} \right)(ax) d\mu(x) \right)$$

$$= \int \varphi(x) \left(\sum_{a \in S} \frac{1}{\sigma(a)} f(ax) g_\mu(ax) \right) d\mu(x)$$

$$= \int \varphi(x) \left(\sum_{\sigma y = x} g_\mu(y) f(y) \right) d\mu(x).$$

Since this holds for all $\varphi \in L^\infty$, $\hat{T}_\mu f = \sum_{\sigma y = x} g_\mu(y) f(y)$ in L^1 . \square

Properties of the Transfer Operator Let μ be a

σ -inv prob measure, then:

$$(1) \forall \varphi \in L^\infty(\mu), f \in L^1(\mu), \quad \int \varphi \hat{T}_\mu f \, d\mu = \int \varphi \circ \sigma f \, d\mu$$

$$(2) \quad E_\mu(f | \mathcal{A}_1^\infty) = (\hat{T}_\mu f) \circ \sigma, \quad \mathcal{A}_1^\infty := \{[a] : a \in S\},$$

$$\mathcal{A}_1^\infty := \left(\begin{array}{l} \sigma\text{-algebra generated} \\ \text{by } \bigcup_{i=1}^\infty \sigma^{-i} \mathcal{A} \end{array} \right)$$

Proof: The first part follows from the previous proof.

To see the second part we recall the defⁿ of the definition of the conditional expectation $E_\mu(f | \mathcal{A}_1^\infty)$: It's the unique $L^1(\mu)$ element s.t.

(a) $E(f | \mathcal{A}_1^\infty)$ is \mathcal{A}_1^∞ -measurable;

(b) $\forall \varphi \in L^\infty$ \mathcal{A}_1^∞ -measurable, $\int \varphi E(f | \mathcal{A}_1^\infty) \, d\mu = \int \varphi f \, d\mu$.

We need the following exercise: $\varphi \in L^\infty$ is \mathcal{A}_1^∞ -measurable iff $\varphi = \psi \circ \sigma$ where ψ is Borel.

Exercise in Measure Theory: Suppose φ is bounded, then φ is \mathcal{A}_1^∞ -measurable iff $\varphi = \psi \circ \sigma$ for ψ Borel.

By the exercise,

(a) $(\hat{T}_\mu f) \circ \sigma$ is \mathcal{A}_1^∞ -measurable, and

(b) $\forall \varphi \in L^\infty$ \mathcal{A}_1^∞ -meas, $\varphi = \psi \circ \sigma$ with ψ Borel meas, and

$$\begin{aligned} \int \varphi (\hat{T}_\mu f) \circ \sigma \, d\mu &= \int \psi \circ \sigma (\hat{T}_\mu f) \circ \sigma \, d\mu = \int \psi \hat{T}_\mu f \, d\mu \\ &= \int \psi \circ \sigma \cdot f \, d\mu = \int \varphi f \, d\mu. \end{aligned}$$

So $(\hat{T}_\mu f) \circ \sigma = E_\mu(f | \mathcal{A}_1^\infty)$ a.e.

Solution of Exercise: The Borel σ -algebra \mathcal{B} is generated by the cylinder, therefore $\mathcal{B} = \mathcal{A}_0^\infty$. It follows that $\sigma^{-1}\mathcal{B} = \mathcal{A}_1^\infty$. Thus φ is \mathcal{A}_1^∞ -measurable iff φ is $\sigma^{-1}\mathcal{B}$ -measurable.

Any $\varphi = \psi \circ \sigma$ with ψ Borel is $\sigma^{-1}(\mathcal{B})$ -measurable:

$$\{x: \varphi(x) > t\} = \{x: \psi(\sigma x) > t\} = \sigma^{-1}\{y: \psi(y) > t\} \in \mathcal{B}.$$

Conversely, if φ is $\sigma^{-1}\mathcal{B}$ -measurable, then $\forall |k| < N+1$

$\{x: \frac{k}{N} \leq \varphi(x) < \frac{k+1}{N}\} \in \sigma^{-1}\mathcal{B}$, therefore $\exists A_{k,N} \in \mathcal{B}$ s.t.

$$\{x: \frac{k}{N} \leq \varphi(x) < \frac{k+1}{N}\} = \sigma^{-1}A_{k,N}, \text{ whence}$$

$$\varphi(x) \equiv \lim_{N \rightarrow \infty} \sum_{k=-\lfloor \varphi \rfloor_{N-1}}^{\lfloor \varphi \rfloor_{N+1}} \frac{k}{N} 1_{\left[\frac{k}{N} \leq \varphi < \frac{k+1}{N}\right]} \quad (\text{uniform convergence})$$

$$= \lim_{N \rightarrow \infty} \sum_{k=-\lfloor \varphi \rfloor_{N-1}}^{\lfloor \varphi \rfloor_{N+1}} \frac{k}{N} 1_{A_{k,N}} \circ \sigma = \psi \circ \sigma$$

$$\text{for the Borel function } \psi := \lim_{N \rightarrow \infty} \sum_{k=-\lfloor \varphi \rfloor_{N-1}}^{\lfloor \varphi \rfloor_{N+1}} \frac{k}{N} 1_{A_{k,N}}.$$

(The previous formula shows that the limit exists everywhere) \square

Entropy Formula: Our aim is to show

Thm. Suppose m_ϕ is the equilibrium measure of a (bounded) Hölder continuous potential on a Markov shift with finite top entropy. Then:

$$h_{m_\phi}(\sigma) = - \int \log g_{m_\phi} dm_\phi$$

(Here $g_{m_\phi} = dm_\phi / d\sigma$ is the Jacobian of m_ϕ .)

Review of Entropy Theory: Let m be a σ -inv prob measure.

- Natural Partition: $\alpha = \{[a] : a \in S\}$
- Entropy of α : $H_m(\alpha) = - \sum_{A \in \alpha} m(A) \log m(A)$
- Information Function: $I_m(\alpha | \alpha_i^\infty) = - \sum_{a \in S} 1_{[a]} \log \mathbb{E}_m(1_{[a]} | \alpha_i^\infty)$
- Conditional Entropy: $H_m(\alpha | \alpha_i^\infty) = \int I_m(\alpha | \alpha_i^\infty) dm$
- Sinai-Rokhlin Thm: If $H_m(\alpha) < \infty$, then

$$h_m(\sigma) \stackrel{!}{=} h_m(\sigma, \alpha) \stackrel{!}{=} H_m(\alpha | \alpha_i^\infty) \equiv \int I_m(\alpha | \alpha_i^\infty) dm$$

(The marked identification could be false when $H_m(\alpha) = \infty$.)

Ledrappier Identity: For any inv Borel prob measure on a Markov shift,

$$I_m(\alpha | \alpha_i^\infty) = - \log g_m$$

Exercise: Use the transfer operator to prove Ledrappier's identity.

Solution:
$$I_m(\alpha | \alpha_i^\infty)(x) = - \sum_{a \in S} 1_{[a]}(x) \log IE_m(1_{[a]} | \alpha_i^\infty)(x)$$

$$= - \sum_{a \in S} 1_{[a]}(x) \log \left(\bigwedge_m 1_{[a]} \right)(\sigma x) \quad (\because IE_m(f | \alpha_i^\infty) = \left(\bigwedge_m f \right) \circ \sigma)$$

$$= - \sum_{a \in S} 1_{[a]}(x) \log \sum_{\sigma y = \sigma x} g_m(y) 1_{[a]}(y)$$

$$= - \log \sum_{\sigma y = \sigma x} g_m(y) 1_{[x_0]}(y) \quad (\because a \neq x_0 \Rightarrow 1_{[a]}(x) = 0)$$

$$= - \log g_m(x), \text{ because } \left. \begin{matrix} \sigma y = x_0 \\ y \in [x_0] \end{matrix} \right\} \Rightarrow y = x. \quad \square$$

Proof of Entropy Formula: If $H_{m_\phi}(\alpha) < \infty$, then the Sinai-Rokhlin theory tells us that

$$h_{m_\phi}(\sigma) = \int I_{m_\phi}(\alpha | \alpha_i^\infty) dm_\phi = - \int \log g_{m_\phi} dm_\phi.$$

\uparrow Sinai-Rokhlin \uparrow Ledrappier

The rest of the proof treats the case $H_{m_\phi}(\alpha) = \infty$.

The main idea is to consider the induced map on some partition set $[a]$, and show that it's a Markov shift, and that its natural partition has finite entropy for the "induced" measure $\bar{m}_\phi = m_\phi(\cdot | [a])$. Now

$$\left(\begin{matrix} \text{Rokhlin's formula} \\ \text{for the induced map} \end{matrix} \right) + \left(\begin{matrix} \text{Abramov's} \\ \text{formula} \end{matrix} \right) + \left(\begin{matrix} \text{Kac} \\ \text{formula} \end{matrix} \right) \Rightarrow \left(\begin{matrix} \text{Rokhlin's} \\ \text{formula for} \\ m_\phi \end{matrix} \right)$$

For details, see:

Buzzi, Sarig: Uniqueness of equilibrium measures for countable Markov shifts and multidimensional piecewise expanding maps. *Ergodic Theory & Dynamical Syst.* 23, 1383-1400 (2005). \square

Proof of the Structure Theorem (1): Suppose $\phi: \Sigma^+ \rightarrow \mathbb{R}$

is a (bounded) Hölder function on a top transitive Markov shift Σ^+ with finite top entropy. Necessarily,

$$P_\sigma(\phi) = \sup \left\{ h_m(\sigma) + \int \phi dm \right\} < \infty.$$

Let $\lambda = \exp P_\sigma(\phi)$. Recall that $\exists h \geq 0$ continuous s.t. $L_\phi h \leq \lambda h$. Let

$$g(x) := \frac{e^{\phi(x)} h(x)}{\lambda h(\sigma x)}$$

Exercise: For each x

$$(1) \sum_{\sigma y = x} g(y) \leq 1,$$

$$(2) 0 < g(x) \leq 1,$$

$$(3)^* \log g \in L^1(m_\phi) \text{ for any equilibrium measure } m_\phi,$$

$$(4) \log h - \log h \circ \sigma \in L^1(m_\phi), \text{ therefore } \int (\log h - \log h \circ \sigma) dm_\phi = 0.$$

(Solⁿ of (3)): See [Burzi - Sarig, ETDS (2003)], p. 1391).

Exercise: Suppose m is an inv prob measure with finite entropy.

Let $g_m = \log \frac{dm}{dm \circ \sigma}$. Then:

$$(1) \sum_{\sigma y = x} g_m(y) = 1 \quad m\text{-a.e. (Hint: Prove that } \hat{\tau}_m 1 = 1)$$

$$(2) 0 \leq g_m(y) \leq 1 \quad m\text{-a.e.}$$

$$(3) \text{ If } m \text{ satisfies Rokhlin's Entropy formula, then } g_m \in L^1(m)$$

Now suppose m_ϕ is an equilibrium measure of ϕ . Then

$$0 = h_{m_\phi}(\sigma) + \int \phi dm_\phi - P_\phi(\phi)$$

prev. exercise \nearrow

$$\stackrel{!}{=} h_{m_\phi}(\sigma) + \int [\phi + (\log h - \log h \circ \sigma) - P_\phi(\phi)] dm_\phi$$

$$= h_{m_\phi}(\sigma) + \int \log g dm_\phi$$

Rohlin formula \nearrow

$$\stackrel{!}{=} - \int \log g_{m_\phi} dm_\phi + \int \log g dm_\phi = \int \log \left(\frac{g}{g_{m_\phi}} \right) dm_\phi$$

$$= \int \hat{T}_{m_\phi} \left[\log \frac{g}{g_{m_\phi}} \right] dm_\phi \quad \left(\because \int \hat{T}_m \varphi dm = \int 1 \circ \sigma \cdot \varphi dm = \int \varphi dm \right)$$

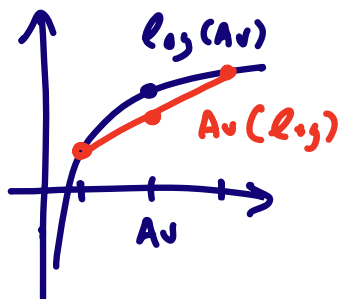
$$= \int \underbrace{\sum_{\sigma y=x} g_{m_\phi}(y) \log \frac{g(y)}{g_{m_\phi}(y)}}_{\text{weighted average}} dm_\phi$$

weighted average
Since $\sum_{\sigma y=x} g_{m_\phi}(y) = 1$

\leq \nearrow

$$\int \log \left(\sum_{\sigma y=x} g_{m_\phi}(y) \frac{g(y)}{g_{m_\phi}(y)} \right) dm_\phi = \int \log \underbrace{\sum_{\sigma y=x} g(y)}_{\leq 1} dm_\phi \leq 0.$$

Jensen's Inequality for the concave $\log x$



Necessarily the red ineq. is an equality. \Downarrow strict concavity of $\log \cdot$

For m_ϕ -a.e. x , $\frac{g(y)}{g_{m_\phi}(y)}$ are equal for all $y \in \sigma^{-1}(x)$ s.t. $g_{m_\phi}(y) \neq 0$

Equivalently, $\exists c(x)$ s.t. $g(y) = c(x) g_{m_\phi}(y) \forall y \in \sigma^{-1}(x)$ s.t. $g_{m_\phi}(y) \neq 0$.

Substituting this above, we obtain $\int \log(c(x) \underbrace{\sum_{\sigma y=x} g_{m_\phi}(y)}_{=1 \text{ a.e.}}) dm_\phi = 0$

So $\int \log c(x) dm_\phi = 0$.

Also $1 \geq \sum_{\sigma y=x} g(y) \geq c(x) \sum_{\sigma y=x} g_{m_\phi}(y) = c(x) \text{ } m_\phi\text{-a.e.}$, so $c(x) \leq 1 \text{ a.e.}$

Necessarily $c(x) = 1 \text{ a.e.}$, and so

$$\boxed{\text{For } m_\phi\text{-a.e. } x, \left. \begin{array}{l} \sigma y=x \\ g_{m_\phi}(y) > 0 \end{array} \right\} \Rightarrow g_{m_\phi}(y) = g(x)}$$

Thus: $(\hat{T}_{m_\phi} f)(x) = \sum_{\sigma y=x} g(y) f(y) = \lambda^{-1} h^{-1} L_\phi(hf)$

and we have identified the transfer operator of m_ϕ .

We are now nearly done: Let $\nu = h^{-1} dm_\phi$. Then:

(1) $dm_\phi = h d\nu$: Trivial

(2) $L_\phi^* \nu = \lambda \nu$: $\forall f \in L^1(\nu)$, $(L_\phi^* \nu)(f) = \int (L_\phi f) d\nu = \int h^{-1} L_\phi(h \frac{f}{h}) d\nu$
 $= \lambda \int \hat{T}_{m_\phi}(f/h) dm_\phi = \lambda \int (f/h) dm_\phi = \lambda \int f d\nu = \lambda \nu(f).$

$$\Rightarrow L_\phi^* \nu = \lambda \nu.$$

Exercise: Show that $\text{supp}(\nu) = \Sigma^+$. (Hint: Use transitivity to show that \forall cylinders $[a], [b] \exists N$ s.t. $L_\phi^N 1_{[a]} > 0$ on $[b]$).

(3) $L_\phi h = \lambda h$: Since m_ϕ is σ -inv, $\hat{T}_{m_\phi} 1 = 1$, whence
 $\lambda^{-1} h^{-1} L_\phi(h \cdot 1) = \hat{T}_{m_\phi} 1 = 1 \Rightarrow L_\phi h = \lambda h \text{ } m_\phi\text{-a.e.}$
 $\Rightarrow L_\phi h = \lambda h$ everywhere (because both sides of the equation are continuous functions, and $\text{supp}(m_\phi) = \text{supp}(\nu) = \Sigma^+$). \square

Proof that m_ϕ is Bernoulli - up to - a Period : We

do this in the special case of MMEs ($\phi \equiv 0$).

The general case can be found in §3 of :

O. Sarig : Bernoulli equilibrium states for diffeomorphisms,
J. Modern Dynamics (2011)

Recall that if $\phi \equiv 0$, then h is constant on partition sets,
i.e. $h = h(x_0)$. Therefore,

$$\begin{aligned} m_\phi[a_0, \dots, a_n] &= m_\phi(1_{[a_0, \dots, a_n]}) = m_\phi(\hat{T}_{m_\phi}^n 1_{[a_0, \dots, a_n]}) \\ &= m_\phi(\lambda^{-n} h^{-1} L_\phi^n (h 1_{[a_0, \dots, a_n]})) \quad (\text{check!}) \\ &= m_\phi(\lambda^{-n} h^{-1} \sum_{\phi^n y = x} e^{\phi^n(y)} h(y) 1_{[a_0, \dots, a_n]}(y)) \\ &= m_\phi(\lambda^{-n} h^{-1} h(a_0) \underbrace{1_{[a_0, \dots, a_n]}(a_0, \dots, a_{n-1}, x)}_{\equiv 1_{[a_n]}(x)}) \\ &\stackrel{\phi \equiv 0}{=} m_\phi(\lambda^{-n} \frac{h(a_0)}{h(a_n)} 1_{[a_n]}) \quad \equiv 1_{[a_n]}(x) \\ &= \frac{m_\phi[a_n] h(a_0)}{\lambda^n h(a_n)} = m_\phi[a_0] \cdot \frac{h(a_0)/m_\phi[a_0]}{\lambda^n h(a_n)/m_\phi[a_n]} \\ &\stackrel{h(x)=h(y)}{=} m_\phi[a_0] \frac{1/\partial[a_0]}{1/\partial[a_n]} = m_\phi[a_0] \cdot \frac{\partial[a_n]}{\lambda^n \partial[a_0]} = m_\phi[a_0] \cdot \prod_{i=0}^n \frac{\partial[a_{i+1}]}{\lambda \partial[a_i]} \\ &= p_{a_0} \cdot p_{a_0 a_1} \cdot \dots \cdot p_{a_{n-1} a_n}, \text{ where} \end{aligned}$$

$p_a = m_\phi(a) \text{ and } p_{a,b} = \begin{cases} \frac{\partial(b)}{\lambda \partial(a)} & a \rightarrow b \\ 0 & \text{else.} \end{cases}$
--

Exercise: (p_a) is a prob vector; $(p_{a,b})$ is a stoch. matrix, and $\forall a \ p_a \neq 0$; If $a \rightarrow b$, then $p_{a,b} \neq 0$.

[Solⁿ: (p_a) is a prob. vector because m_ϕ is a prob. measure. Next, by the previous formula $p_{ab} = \frac{p_a p_{ab}}{p_a} = \frac{m_\phi(c,b)}{m_\phi(c)}$, so $\sum_{b: a \rightarrow b} p_{ab} = \sum_{b: a \rightarrow b} \frac{m_\phi(c,b)}{m_\phi(c)} = \frac{m_\phi(c)}{m_\phi(c)} = 1$.]

The exercise proves:

Gurevich's Thm: The MME of a top transitive Markov shift is a globally supported Markov measure.

It's well-known that globally supported Markov measures on transitive Markov shifts are Bernoulli-up-to-a-period. \square

Uniqueness of the Equilibrium Measure: Suppose by way of contradiction there were two different equilibrium measures m'_ϕ, m''_ϕ . Then $m_\phi := \frac{1}{2}(m'_\phi + m''_\phi)$ is a non-ergodic equilibrium measure. But we just saw that eq. measures are Bernoulli up to a period, and this implies ergodicity. \square

References for Lecture 2

Main Coding Theorem: We use the formulation in

Buzzi - Craisier - Sarig: Strong Positive Recurrence and Exponential Mixing for Diffeomorphisms. ArXiv (2025).

See also:

Sarig: Symbolic dynamics for surface diffeomorphisms with positive entropy, J. of AMS (2018)

Ben Ovadia: Symbolic dynamics for non-uniformly hyperbolic diffeomorphisms of compact smooth manifolds. J. Modern Dynam (2018)

Buzzi - Craisier - Sarig: Measures of maximal entropy for surface diffeomorphisms, Ann. Math. (2022)

Araujo - Lima - Poletti: Symbolic dynamics for non uniformly hyperbolic maps with singularities in higher dimension. Memoirs of AMS (2024)

Structure of MME for Markov Shifts:

Buzzi & Sarig: Uniqueness of equilibrium measures for countable Markov shifts and multidimensional piecewise expanding maps ETDS (2003)

Gurevich: Shift entropy and Markov measures in the space of paths of a countable graph. Dokl. Acad. Nauk. SSSR (1970)

Parry: Intrinsic Markov chains. Trans. AMS (1964)

Sarig: Thermodynamic formalism for countable Markov shifts. Lecture notes freely available from my homepage.

Bernoulli Property: **Sarig**: Bernoulli equilibrium states for surface diffeomorphisms. J. Modern Dynam. (2011)