

ERGODIC MEASURES

In this chapter we shall consider the stronger property of ergodicity for an invariant probability measure μ . This property is more appropriate (amongst other things) for understanding the “long term” average behaviour of a transformation.

9.1 Definitions and characterization of ergodic measures

DEFINITION. Given a probability space (X, \mathcal{B}, μ) , a transformation $T : X \rightarrow X$ is called *ergodic* if for every set $B \in \mathcal{B}$ with $T^{-1}B = B$ we have that either $\mu(B) = 0$ or $\mu(B) = 1$.

Alternatively we say that μ is T -ergodic.

The following lemma gives a simple characterization in terms of functions.

LEMMA 9.1. *T is ergodic with respect to μ iff whenever $f \in L^1(X, \mathcal{B}, \mu)$ satisfies $f = f \circ T$ then f is a constant function.*

PROOF. This is an easy observation using indicator functions. ■

9.2 Poincaré recurrence and Kac's theorem

We begin with one of the most fundamental results in ergodic theory.

THEOREM 9.2 (POINCARÉ RECURRENCE THEOREM). *Let $T : X \rightarrow X$ be a measurable transformation on a probability space (X, \mathcal{B}, μ) . Let $A \in \mathcal{B}$ have $\mu(A) > 0$; then for almost points $x \in A$ the orbit $\{T^n x\}_{n \geq 0}$ returns to A infinitely often.*

PROOF. Let $F = \{x \in A : T^n x \notin A, \forall n \geq 1\}$, then it suffices to show that $\mu(F) = 0$.

Towards this end, we first observe that $T^{-m}F \cap T^{-n}F = \emptyset$ when $n > m$, say. If this were not the case and $w \in T^{-m}F \cap T^{-n}F$ then $T^m w \in F$ and $T^{n-m}(T^m w) \in F \subset A$, which contradicts the definition of F .

Thus since the sets $\{T^{-n}F\}_{n \geq 0}$ are disjoint we see that

$$\sum_{n=0}^{\infty} \mu(T^{-n}F) = \mu(\cup_{n=0}^{\infty} T^{-n}F) \leq \mu(X) = 1$$

and then because μ is T -invariant $\mu(F) = \mu(T^{-1}F) = \dots = \mu(T^{-n}F) = \dots$ so we can only have that $\mu(F) = 0$. ■

DEFINITION. Let $n_A : A \rightarrow \mathbb{Z}^+ \cup \{+\infty\}$ be the first return time i.e. $n_A(x) > 0$ is the smallest value for which $T^{n_A(x)}x \in A$.

By Theorem 9.2 $n_A(x)$ is finite almost everywhere. The next theorem shows that when μ is an ergodic measure then the *average* return time to A can be calculated explicitly.

THEOREM 9.3 (KAC'S THEOREM). *Let $T : X \rightarrow X$ be an ergodic transformation on a probability space (X, \mathcal{B}, μ) . Let $A \in \mathcal{B}$ have $\mu(A) > 0$ then we define the return time function $n_A : A \rightarrow \mathbb{Z}^+ \cup \{\infty\}$ (which is finite, almost everywhere). The average return time (with respect to the induced probability measure μ_A) is*

$$\int_A n_A(x) d\mu_A(x) = \frac{1}{\mu(A)}.$$

PROOF. By definition of μ_A it is equivalent to show that $\int_A n_A(x) d\mu(x) = 1$. It is useful to define the following sets.

- (a) For each $n \geq 1$ we define $A_n = \{x \in A : n_A(x) = n\}$, and write $A = \cup_{n \geq 1} A_n$ (with $A_i \cap A_j = \emptyset$ for $i \neq j$). In particular, $\sum_{n=1}^{\infty} \mu(A_n) = \mu(A)$.
- (b) For $n \geq 1$ we define $B_n = \{x \in X : T^j x \notin A \text{ for } 1 \leq j \leq n-1, T^n x \in A\}$. The sets B_n are disjoint (i.e. $B_i \cap B_j = \emptyset$ for $i \neq j$) and by ergodicity $X = \cup_{n \geq 1} B_n$ (since $\cup_{n \geq 1} B_n \supset \cup_{n \geq 1} T^{-n}A \supset A$) so that $\sum_{n=1}^{\infty} \mu(B_n) = 1$.

We can rewrite

$$\int_A n_A(x) d\mu(x) = \sum_{k=1}^{\infty} k \mu(A_k) = \sum_{k=1}^{\infty} \left(\sum_{n=k}^{\infty} \mu(A_n) \right);$$

then if we can show that $\sum_{n=k}^{\infty} \mu(A_n) = \mu(B_k)$ this will complete the proof.

When $k = 1$ we have from the definitions that $B_1 = T^{-1}A$ and so $\sum_{n=1}^{\infty} \mu(A_n) = \mu(A) = \mu(B_1)$, as required. For $k > 1$ we can proceed by induction. We can partition $T^{-1}B_k = B_{k+1} \cup T^{-1}A_k$ (where $T^{-1}A_k = T^{-1}B_k \cap T^{-1}A$ and $B_{k+1} = T^{-1}B_k \cap (X - T^{-1}A)$). Thus $\mu(T^{-1}B_k) = \mu(B_k) = \mu(B_{k+1}) + \mu(T^{-1}A_k) = \mu(B_{k+1}) + \mu(A_k)$ and using the inductive hypothesis we see that $\mu(B_{k+1}) = \mu(B_k) - \mu(A_k) = \sum_{n=k+1}^{\infty} \mu(A_n)$. This completes the inductive step and the proof. ■

9.3 Existence of ergodic measures

When $T : X \rightarrow X$ is a continuous map on a compact metric space there is a very simple relationship between ergodic measures and invariant measures which we can now describe.

Let \mathcal{M} denote the set of invariant probability measures on X . There is a natural topology on this space called the *weak-star* topology, i.e. the weakest topology such that a sequence $\mu_n \in \mathcal{M}$ converges to $\mu \in \mathcal{M}$ iff $\forall f \in C^0(X)$, $\int f d\mu_n \rightarrow \int f d\mu$.

The following properties of \mathcal{M} are well-known (and easily checked):

- (i) \mathcal{M} is convex (i.e. if $\mu_1, \mu_2 \in \mathcal{M}$ and $0 < \alpha < 1$, then $\alpha\mu_1 + (1-\alpha)\mu_2 \in \mathcal{M}$);
- (ii) the set \mathcal{M} is compact (in the weak-star topology) [5, Theorem 6.10].

LEMMA 9.4. *The extremal points in the convex set \mathcal{M} are ergodic measures (i.e. $\mu \in \mathcal{M}$ is ergodic if whenever $\exists \mu_1, \mu_2 \in \mathcal{M}$ and $0 < \alpha < 1$ with $\mu = \alpha\mu_1 + (1-\alpha)\mu_2$ then $\mu_1 = \mu_2$).*

The converse is also true, but we shall not require it.

PROOF. If μ is not ergodic then we can find $B \in \mathcal{B}$ with $T^{-1}B = B$ and $0 < \mu(B) < 1$. But for any set $A \in \mathcal{B}$ we can write $A = (A \cap B) \cup (A \cap (X - B))$ and thus

$$\begin{aligned} \mu(A) &= \mu((A \cap B) \cup (A \cap (X - B))) \\ &= \mu(B) \left(\frac{\mu(A \cap B)}{\mu(B)} \right) + \mu(X - B) \left(\frac{\mu(A \cap (X - B))}{\mu(X - B)} \right) \\ &= \alpha\mu_1(A) + (1 - \alpha)\mu_2(A) \end{aligned}$$

where $\alpha = \mu(B)$ and $\mu_1(A) = \frac{\mu(A \cap B)}{\mu(B)}$, $\mu_2(A) = \frac{\mu(A \cap (X - B))}{\mu(X - B)}$. This shows that $\mu = \alpha\mu_1 + (1 - \alpha)\mu_2$. ■

PROPOSITION 9.5 (EXISTENCE OF ERGODIC MEASURES). *Let X be a compact metric space and \mathcal{B} be the Borel sigma-algebra. Given any continuous map $T : X \rightarrow X$ there exists at least one T -ergodic probability measure μ .*

PROOF. Choose a dense set of functions $f_k \in C^0(X)$, $k \geq 0$. Since the map $\mu \rightarrow \int f_0 d\mu$ is continuous on \mathcal{M} there exists by (weak-star) compactness at least one $\nu \in \mathcal{M}$ such that $\int f_0 d\nu = \sup_{\mu \in \mathcal{M}} \{ \int f_0 d\mu \}$. We let

$$\mathcal{M}_0 = \left\{ \nu \in \mathcal{M} : \int f_0 d\nu = \sup_{\mu \in \mathcal{M}} \left\{ \int f_0 d\mu \right\} \right\};$$

then clearly \mathcal{M}_0 is non-empty and closed. Similarly, define

$$\mathcal{M}_1 = \left\{ \nu \in \mathcal{M}_0 : \int f_1 d\nu = \sup_{\mu \in \mathcal{M}_0} \left\{ \int f_1 d\mu \right\} \right\}$$

and the same reasoning shows that $\mathcal{M}_1 \subset \mathcal{M}_0 \subset \mathcal{M}$ is non-empty and closed.

Proceeding inductively we define

$$\mathcal{M}_k = \left\{ \nu \in \mathcal{M}_{k-1} : \int f_k d\nu = \sup_{\mu \in \mathcal{M}_{k-1}} \left\{ \int f_k d\mu \right\} \right\}$$

and arrive at a nested sequence $\mathcal{M} \supset \mathcal{M}_0 \supset \mathcal{M}_1 \supset \mathcal{M}_2 \supset \dots \supset \mathcal{M}_k \supset \dots$. Since the sets are all closed in \mathcal{M} (and hence compact) we have that the intersection is non-empty. Assume $\mu \in \bigcap_{k \in \mathbb{Z}^+} \mathcal{M}_k$. We want to show that μ is ergodic by showing that it is an extreme point in \mathcal{M} .

Assume that μ can be written as an affine combination $\mu = \alpha\mu_1 + (1-\alpha)\mu_2$ (with $0 < \alpha < 1$); then to show that μ is ergodic we need to show that $\mu_1 = \mu_2$. Thus it suffices to show that for every $f_k \in C^0(X)$ we have that $\int f_k f d\mu_1 = \int f_k f d\mu_2$ (since the set f_k is dense).

We begin with $k = 0$ and observe that by assumption $\int f_0 d\mu = \alpha \int f_0 d\mu_1 + (1-\alpha) \int f_0 d\mu_2$. Since $\mu \in \mathcal{M}_0$ we see that $\sup_{m \in \mathcal{M}} \{ \int f_0 dm \} = \int f_0 d\mu$ implies that $\int f_0 d\mu_1 = \int f_0 d\mu_2 = \sup_{m \in \mathcal{M}} \{ \int f_0 dm \}$. We thus conclude

- (1) the first identity $\int f_0 d\mu_1 = \int f_0 d\mu_2$ is proved.
- (2) $\mu_1, \mu_2 \in \mathcal{M}_0$.

Continuing inductively, we establish that for arbitrary $k \geq 0$ we have $\int f_k d\mu_1 = \int f_k d\mu_2$ and $\mu_1, \mu_2 \in \mathcal{M}_k$. This completes the proof (i.e. $\mu_1 = \mu_2$ and μ is an extremal measure). ■

REMARK. The following facts are easy to check.

- (3) If ν, μ are distinct T -ergodic measures then $\nu \perp \mu$.
- (4) If μ is ergodic then it is an extremal measure in \mathcal{M} . (The converse to Lemma 9.4.)

Since \mathcal{M} is a compact convex metric space there is a general theorem of Choquet that says *every invariant measure $\mu \in \mathcal{M}$ can be written as a convex combination of extremal measures in \mathcal{M}* . More precisely, we can find a measure $\rho = \rho_\mu$ on the space \mathcal{M} (with respect to the Borel sigma-algebra associated to the weak-star topology) such that

- (1) for any function $f \in C^0(X)$ we have

$$\int f d\mu = \int_{\mathcal{M}} \left(\int f d\nu \right) d\rho(\nu).$$

- (2) $\rho(\{\nu : \nu \text{ is extremal}\}) = 1$.

9.4 Some basic constructions in ergodic theory

In this final section of chapter 9 we shall describe two basic constructions in ergodic theory.

9.4.1. Skew products. Let $T : X \rightarrow X$ be a measure preserving transformation of a probability space (X, \mathcal{B}, μ) . Let (G, \mathcal{B}) be a compact Lie group with the Borel sigma-algebra \mathcal{B} . We can consider the product space $X \times G$ with the product sigma-algebra \mathcal{A} .

DEFINITION. Given a measure preserving transformation of $T : X \rightarrow X$ and a measurable map $\phi : X \rightarrow G$ we define a *skew product* to be the transformation $S : X \times G \rightarrow X \times G$ defined by $S(x, g) = (Tx, \phi(x)g)$. Given any T -invariant probability measure μ we can associate the S -invariant measure ν defined by $d\nu = d\mu \times dt$.

A simple example is the following.

EXAMPLE. Let $T : \mathbb{R}/\mathbb{Z} \rightarrow \mathbb{R}/\mathbb{Z}$ be given by $T(x) = x + \alpha \pmod{1}$ for some $\alpha \in \mathbb{R}$. Let $G = \mathbb{R}/\mathbb{Z}$ and we define $\phi : \mathbb{R}/\mathbb{Z} \rightarrow \mathbb{R}/\mathbb{Z}$ by $\phi(x) = x \pmod{1}$ (i.e. the identity map). The associated skew product is then the map $S : \mathbb{R}^2/\mathbb{Z}^2 \rightarrow \mathbb{R}^2/\mathbb{Z}^2$ given by $S(x, y) = (x + \alpha, x + y) \pmod{1}$.

9.4.2. Induced transformations and Rohlin towers. Assume that $T : X \rightarrow X$ is a measurable transformation on a measurable space (X, \mathcal{B}) . Assume that $A \subset X$ with $A \in \mathcal{B}$.

DEFINITION. The transformation $T_A : A \rightarrow A$ defined by $T_A(x) = T^{n_A(x)}x$ is called the *induced transformation* on A . We denote by $\mathcal{B}_A = \{B \cap A : B \in \mathcal{B}\}$ the restriction of the sigma-algebra \mathcal{B} to A .

If μ is a T -invariant sigma-finite measure on (X, \mathcal{B}) and $0 < \mu(A) < \infty$ then we can define a T_A -invariant measure μ_A on (A, \mathcal{B}_A) by $\mu_A(B) = \frac{\mu(A \cap B)}{\mu(A)}$.

EXAMPLE (CONTINUED FRACTION TRANSFORMATION). Consider the case where (X, \mathcal{B}) is the positive half-line $\mathbb{R}^+ = (0, +\infty)$ with the Borel sigma-algebra. We define a transformation $T : \mathbb{R}^+ \rightarrow \mathbb{R}^+$ by

- (1) $Tx = x - 1$ if $x \in [1, +\infty)$, and
- (2) $Tx = \frac{1}{x}$ if $x \in (0, 1)$.

We can consider the induced transformation $T_A : A \rightarrow A$ on the interval $A = (0, 1]$ defined by $T_Ax = \frac{1}{x} - [\frac{1}{x}]$.

The measure μ_A defined by $\mu_A(B) = \frac{1}{\log 2} \int_B \frac{1}{1+x} dx$ is T_A -invariant.

REMARK. We need not be too careful about the definition of T and T_A on a countable set of points since they have zero measure.

Consider an (ergodic) transformation $T : X \rightarrow X$ on a probability space (X, \mathcal{B}, μ) and let $A \in \mathcal{B}$ have $\mu(A) > 0$.

DEFINITION. We can define a space

$$A^{n_A} = \{(x, k) \in A \times \mathbb{Z}^+ : 0 \leq k \leq n_A(x)\},$$

where we identify $(x, n_A(x)) \sim (T^{n_A(x)}x, 0)$, and introduce the product sigma-algebra $\bar{\mathcal{B}}$ (i.e. the smallest sigma-algebra containing the products of sets in \mathcal{B}_A and $\mathcal{B}_{\mathbb{Z}^+}$).

We define a probability measure on the space A^{n_A} by $\nu = \frac{\mu \times dn}{\int n_A d\mu}$ (where dn corresponds to the usual counting measure on \mathbb{Z}^+).

Finally, we define a transformation $T_A^{n_A} : A^{n_A} \rightarrow A^{n_A}$ by

- (1) $T_A^{n_A}(x, k) = (x, k + 1)$ if $0 \leq k < n_A(x)$, and
- (2) $T_A^{n_A}(x, n_A(x)) = T_A^{n_A}(T_A x, 0) = (T_A x, 1)$.

This construction is called the *Rohlin tower* over A .

(N.B. A Rohlin tower is the converse process to induced transformations. We reproduce the original transformation on X from the induced transformation on A .) The following lemma tells us the Rohlin tower is a good model of the original transformation.

LEMMA 9.6. *The map $\phi : (A^{n_A}, \bar{\mathcal{B}}, \nu) \rightarrow (X, \mathcal{B}, \mu)$ defined by $\phi(x, k) = T^k(x)$ is measurable and satisfies the following:*

- (1) ϕ is a bijection (almost everywhere);
- (2) $\forall B \in \mathcal{B}$ we have that $\nu(\phi^{-1}B) = \mu(B)$; and
- (3) $\phi T_A^{n_A} = T\phi$ (almost everywhere).

PROOF. The result follows almost immediately from the definitions. ■

REMARK. The map ϕ is an *isomorphism* which implies that from the point of view of ergodic theory the transformations T and $T_A^{n_A}$ are the same.

9.4.3 Natural extensions. Given a non-invertible map $T : (X, \mathcal{B}, \mu) \rightarrow (X, \mathcal{B}, \mu)$ there is a natural way of associating to it an invertible transformation $\hat{T} : (\hat{X}, \hat{\mathcal{B}}, \hat{\mu}) \rightarrow (\hat{X}, \hat{\mathcal{B}}, \hat{\mu})$ with similar dynamical properties.

We define

$$\hat{X} = \{(x_n)_{n \in \mathbb{Z}^+} \in \prod_{n \in \mathbb{Z}^+} X : T(x_n) = x_{n+1}, n \geq 0\}$$

and associate the sigma-algebra generated by the sets

$$B_m := \{(x_n)_{n \in \mathbb{Z}^+} \in \hat{X} : x_m \in B\} \text{ for } B \in \mathcal{B} \text{ and } m \in \mathbb{Z}^+.$$

We next define a probability measure $\hat{\mu}$ on $\hat{\mathcal{B}}$ by $\hat{\mu}(B_m) = \mu(B)$. Finally, we define the (invertible) transformation $\hat{T} : \hat{X} \rightarrow \hat{X}$ by

$$\hat{T}(x_0, x_1, x_2, \dots) = (Tx_0, x_0, x_1, x_2, \dots).$$

It is easy to see from the construction that \hat{T} is measurable and preserves the probability measure $\hat{\mu}$.

DEFINITION. We call $\hat{T} : \hat{X} \rightarrow \hat{X}$ the *natural extension* of T .

There is a canonical map $\pi : \hat{X} \rightarrow X$ defined by $\pi((x_n)_{n \in \mathbb{Z}^+}) = x_0$. The natural extension \hat{T} has the following properties:

- (i) \hat{T} is an extension of T in the sense that $\pi \circ \hat{T} = T \circ \pi$; and
- (ii) if we denote by $\hat{\mathcal{B}}^+ \subset \hat{\mathcal{B}}$ the sub-sigma-algebra generated by sets $\{\pi^{-1}(B) : B \in \mathcal{B}\}$ then

$$\dots \subset \hat{T}^{-1}\hat{\mathcal{B}}^+ \subset \hat{\mathcal{B}}^+ \subset T\hat{\mathcal{B}}^+ \subset \dots \subset \bigcup_{n \in \mathbb{Z}^+} T^n \hat{\mathcal{B}}^+ = \mathcal{B}.$$

REMARK. In fact, any transformation satisfying (i) and (ii) will be isomorphic to the natural extension as we have defined it above [3].

EXAMPLE (SUBSHIFTS OF FINITE TYPE). Let $\sigma : X_A^+ \rightarrow X_A^+$ be a (one-sided) subshift of finite type, defined by the $k \times k$ matrix A . Relative to a Markov measure, say, its natural extension is the shift $\sigma : X \rightarrow X$.

9.5 Comments and references

More can be found on ergodic measures in [1], [2] and [5].

Important applications of ergodic theory beyond the scope of these notes are Mostow's rigidity theorem [4] and the Margulis super-rigidity theorem [6, §5.1].

The skew product example in subsection 9.4.1 was used by Furstenberg to give a simple proof of a result on diophantine approximation due to Hardy and Littlewood [2].

References

1. P. Billingsley, *Ergodic Theory and Information*, Wiley, New York, 1965.
2. W. Parry, *Topics in Ergodic Theory*, C.U.P., Cambridge, 1981.
3. V. Rohlin, *Exact endomorphisms of lebesgue space*, Amer. Math. Soc. Transl. (2) **39** (1964), 1-36.
4. W. Thurston, *Topology and geometry of three manifolds*, unpublished notes, Princeton University N.J., Princeton, 1978.
5. P. Walters, *An Introduction to Ergodic Theory*, Springer, New York, 1989.
6. R. Zimmer, *Ergodic Theory and Semi-simple Groups*, Birkhäuser, Basel, 1984.