

Introduction to Ergodic Theory and its Applications to Number Theory

Karma Dajani

October 18, 2014

Contents

| | | |
|----------|--|-----------|
| 1 | Introduction and preliminaries | 5 |
| 1.1 | What is Ergodic Theory? | 5 |
| 1.2 | Measure Preserving Transformations | 6 |
| 1.3 | Basic Examples | 10 |
| 1.4 | Recurrence | 16 |
| 1.5 | Induced Transformations | 17 |
| 1.6 | Ergodicity | 20 |
| 1.7 | Other Characterizations of Ergodicity | 22 |
| 1.8 | Examples of Ergodic Transformations | 24 |
| 2 | Ergodic Theorems | 33 |
| 2.1 | The Pointwise Ergodic Theorem and its consequences | 33 |
| 2.2 | The Mean Ergodic Theorem | 46 |
| 2.3 | Mixing | 49 |
| 3 | Measure Preserving Isomorphisms and Factor Maps | 53 |
| 3.1 | Measure Preserving Isomorphisms | 53 |
| 3.2 | Factor Maps | 58 |
| 3.3 | Natural Extensions | 59 |
| 4 | Continued Fractions | 63 |
| 4.1 | Introduction and Basic Properties | 63 |
| 4.2 | Ergodic Properties of Continued Fraction Map | 68 |
| 4.3 | Natural Extension and the Doeblin-Lenstra Conjecture | 72 |
| 5 | Entropy | 77 |
| 5.1 | Randomness and Information | 77 |
| 5.2 | Definitions and Properties | 78 |

| | | |
|----------|--|------------|
| 5.3 | Calculation of Entropy and Examples | 85 |
| 5.4 | The Shannon-McMillan-Breiman Theorem | 89 |
| 5.5 | Lochs' Theorem | 96 |
| 6 | Invariant Measures for Continuous Transformations | 101 |
| 6.1 | Existence | 101 |
| 6.2 | Unique Ergodicity and Uniform Distribution | 108 |
| | Bibliography | 113 |

Chapter 1

Introduction and preliminaries

1.1 What is Ergodic Theory?

It is not easy to give a simple definition of Ergodic Theory because it uses techniques and examples from many fields such as probability theory, statistical mechanics, number theory, vector fields on manifolds, group actions of homogeneous spaces and many more.

The word *ergodic* is a mixture of two Greek words: *ergon* (work) and *odos* (path). The word was introduced by Boltzmann (in statistical mechanics) regarding his hypothesis: *for large systems of interacting particles in equilibrium, the time average along a single trajectory equals the space average*. The hypothesis as it was stated was false, and the investigation for the conditions under which these two quantities are equal lead to the birth of ergodic theory as is known nowadays.

A modern description of what ergodic theory is would be: it is the study of the asymptotic average behavior of systems evolving in time. The collection of all states of the system form a space X , and the evolution is represented by either

- a transformation $T : X \rightarrow X$, where Tx is the state of the system at time $t = 1$, when the system (i.e., at time $t = 0$) was initially in state x . (This is analogous to the setup of discrete time stochastic processes).
- if the evolution is continuous or has a spacial structure, then we describe the evolution by looking at a group of transformations G (like \mathbb{Z}^2 , \mathbb{R} , \mathbb{R}^2) acting on X , i.e., every $g \in G$ is identified with a transformation $T_g : X \rightarrow X$, and $T_{gg'} = T_g \circ T_{g'}$.

The space X usually has a special structure, and we want T to preserve the basic structure on X . For example

- if X is a measure space, then T must be measurable.
- if X is a topological space, then T must be continuous.
- if X has a differentiable structure, then T is a diffeomorphism.

In this course our space is a probability space (X, \mathcal{B}, μ) , and our time is discrete. So the evolution is described by a measurable map $T : X \rightarrow X$, so that $T^{-1}A \in \mathcal{B}$ for all $A \in \mathcal{B}$. For each $x \in X$, the orbit of x is the sequence

$$x, Tx, T^2x, \dots$$

If T is invertible, then one speaks of the two sided orbit

$$\dots, T^{-1}x, x, Tx, \dots$$

We want also that the evolution is in steady state i.e. stationary. In the language of ergodic theory, we want T to be *measure preserving*.

1.2 Measure Preserving Transformations

Definition 1.2.1 Let (X, \mathcal{B}, μ) be a probability space, and $T : X \rightarrow X$ measurable. The map T is said to be *measure preserving with respect to μ* if $\mu(T^{-1}A) = \mu(A)$ for all $A \in \mathcal{B}$.

This definition implies that for any measurable function $f : X \rightarrow \mathbb{R}$, the process

$$f, f \circ T, f \circ T^2, \dots$$

is stationary. This means that for all Borel sets B_1, \dots, B_n , and all integers $r_1 < r_2 < \dots < r_n$, one has for any $k \geq 1$,

$$\begin{aligned} \mu(\{x : f(T^{r_1}x) \in B_1, \dots, f(T^{r_n}x) \in B_n\}) = \\ \mu(\{x : f(T^{r_1+k}x) \in B_1, \dots, f(T^{r_n+k}x) \in B_n\}). \end{aligned}$$

We say T is **invertible** if it is one-to-one and if T^{-1} is measurable. Note that in the case T is invertible, then T is measure preserving if and only if $\mu(TA) = \mu(A)$ for all $A \in \mathcal{B}$. We can generalize the definition of measure preserving to the following case. Let $T : (X_1, \mathcal{B}_1, \mu_1) \rightarrow (X_2, \mathcal{B}_2, \mu_2)$ be measurable, then T is measure preserving if $\mu_1(T^{-1}A) = \mu_2(A)$ for all $A \in \mathcal{B}_2$.

The following gives a useful tool for verifying that a transformation is measure preserving. For this we need the notions of algebra and semi-algebra.

Recall that a collection \mathcal{S} of subsets of X is said to be a *semi-algebra* if (i) $\emptyset \in \mathcal{S}$, (ii) $A \cap B \in \mathcal{S}$ whenever $A, B \in \mathcal{S}$, and (iii) if $A \in \mathcal{S}$, then $X \setminus A = \cup_{i=1}^n E_i$ is a disjoint union of elements of \mathcal{S} . For example if $X = [0, 1)$, and \mathcal{S} is the collection of all subintervals, then \mathcal{S} is a semi-algebra. Or if $X = \{0, 1\}^{\mathbb{Z}}$, then the collection of all cylinder sets $\{x : x_i = a_i, \dots, x_j = a_j\}$ is a semi-algebra.

An *algebra* \mathcal{A} is a collection of subsets of X satisfying: (i) $\emptyset \in \mathcal{A}$, (ii) if $A, B \in \mathcal{A}$, then $A \cap B \in \mathcal{A}$, and finally (iii) if $A \in \mathcal{A}$, then $X \setminus A \in \mathcal{A}$. Clearly an algebra is a semi-algebra. Furthermore, given a semi-algebra \mathcal{S} one can form an algebra by taking all finite disjoint unions of elements of \mathcal{S} . We denote this algebra by $\mathcal{A}(\mathcal{S})$, and we call it the *algebra generated* by \mathcal{S} . It is in fact the smallest algebra containing \mathcal{S} . Likewise, given a semi-algebra \mathcal{S} (or an algebra \mathcal{A}), the σ -algebra generated by \mathcal{S} (\mathcal{A}) is denoted by $\sigma(\mathcal{S})$ ($\sigma(\mathcal{A})$), and is the smallest σ -algebra containing \mathcal{S} (or \mathcal{A}).

A *monotone class* \mathcal{C} is a collection of subsets of X with the following two properties

- if $E_1 \subseteq E_2 \subseteq \dots$ are elements of \mathcal{C} , then $\cup_{i=1}^{\infty} E_i \in \mathcal{C}$,
- if $F_1 \supseteq F_2 \supseteq \dots$ are elements of \mathcal{C} , then $\cap_{i=1}^{\infty} F_i \in \mathcal{C}$.

The *monotone class generated* by a collection \mathcal{S} of subsets of X is the smallest monotone class containing \mathcal{S} .

Theorem 1.2.1 *Let \mathcal{A} be an algebra of X , then the σ -algebra $\sigma(\mathcal{A})$ generated by \mathcal{A} equals the monotone class generated by \mathcal{A} .*

Using the above theorem, one can get an easier criterion for checking that a transformation is measure preserving.

Theorem 1.2.2 *Let $(X_i, \mathcal{B}_i, \mu_i)$ be probability spaces, $i = 1, 2$, and $T : X_1 \rightarrow X_2$ a transformation. Suppose \mathcal{S}_2 is a generating semi-algebra of \mathcal{B}_2 . Then, T is measurable and measure preserving if and only if for each $A \in \mathcal{S}_2$, we have $T^{-1}A \in \mathcal{B}_1$ and $\mu_1(T^{-1}A) = \mu_2(A)$.*

Proof. Let

$$\mathcal{C} = \{B \in \mathcal{B}_2 : T^{-1}B \in \mathcal{B}_1, \text{ and } \mu_1(T^{-1}B) = \mu_2(B)\},$$

then $\mathcal{S}_2 \subseteq \mathcal{C} \subseteq \mathcal{B}_2$, and hence $\mathcal{A}(\mathcal{S}_2) \subset \mathcal{C}$. We show that \mathcal{C} is a monotone class. Let $E_1 \subseteq E_2 \subseteq \dots$ be elements of \mathcal{C} , and let $E = \cup_{i=1}^{\infty} E_i$. Then, $T^{-1}E = \cup_{i=1}^{\infty} T^{-1}E_i \in \mathcal{B}_1$, and

$$\begin{aligned} \mu_1(T^{-1}E) &= \mu_1(\cup_{n=1}^{\infty} T^{-1}E_n) \\ &= \lim_{n \rightarrow \infty} \mu_1(T^{-1}E_n) \\ &= \lim_{n \rightarrow \infty} \mu_2(E_n) \\ &= \mu_2(\cup_{n=1}^{\infty} E_n) \\ &= \mu_2(E). \end{aligned}$$

Thus, $E \in \mathcal{C}$. A similar proof shows that if $F_1 \supseteq F_2 \supseteq \dots$ are elements of \mathcal{C} , then $\cap_{i=1}^{\infty} F_i \in \mathcal{C}$. Hence, \mathcal{C} is a monotone class containing the algebra $\mathcal{A}(\mathcal{S}_2)$. By the monotone class theorem, \mathcal{B}_2 is the smallest monotone class containing $\mathcal{A}(\mathcal{S}_2)$, hence $\mathcal{B}_2 \subseteq \mathcal{C}$. This shows that $\mathcal{B}_2 = \mathcal{C}$, therefore T is measurable and measure preserving. \square

For example if

- $X = [0, 1)$ with the Borel σ -algebra \mathcal{B} , and μ a probability measure on \mathcal{B} . Then a transformation $T : X \rightarrow X$ is measurable and measure preserving if and only if $T^{-1}[a, b) \in \mathcal{B}$ and $\mu(T^{-1}[a, b)) = \mu([a, b))$ for any interval $[a, b)$.
- $X = \{0, 1\}^{\mathbb{N}}$ with product σ -algebra and product measure μ . A transformation $T : X \rightarrow X$ is measurable and measure preserving if and only if

$$T^{-1}(\{x : x_0 = a_0, \dots, x_n = a_n\}) \in \mathcal{B},$$

and

$$\mu(T^{-1}\{x : x_0 = a_0, \dots, x_n = a_n\}) = \mu(\{x : x_0 = a_0, \dots, x_n = a_n\})$$

for any cylinder set.

Exercise 1.2.1 Recall that if A and B are measurable sets, then

$$A \Delta B = (A \cup B) \setminus (A \cap B) = (A \setminus B) \cup (B \setminus A).$$

Show that for any measurable sets A, B, C one has

$$\mu(A \Delta B) \leq \mu(A \Delta C) + \mu(C \Delta B).$$

Another useful lemma is the following; see also [KT].

Lemma 1.2.1 *Let (X, \mathcal{B}, μ) be a probability space, and \mathcal{A} an algebra generating \mathcal{B} . Then, for any $A \in \mathcal{B}$ and any $\epsilon > 0$, there exists $C \in \mathcal{A}$ such that $\mu(A \Delta C) < \epsilon$.*

Proof. First note that since $X \in \mathcal{A}$, then $X \in \mathcal{D}$. Now let $A \in \mathcal{D}$ and $\epsilon > 0$. There exists $C \in \mathcal{A}$ such that $\mu(A \Delta C) < \epsilon$. Since $C^c \in \mathcal{A}$ and $A \Delta C = A^c \Delta C^c$, we have $\mu(A^c \Delta C^c) < \epsilon$ and hence $A^c \in \mathcal{D}$. Finally, suppose $(A_n)_n \subset \mathcal{D}$ and $\epsilon > 0$. For each n , there exists $C_n \in \mathcal{A}$ such that $\mu(A_n \Delta C_n) < \epsilon/2^{n+1}$. It is easy to check that

$$\bigcup_{n=1}^{\infty} A_n \Delta \bigcup_{n=1}^{\infty} C_n \subseteq \bigcup_{n=1}^{\infty} (A_n \Delta C_n),$$

so that

$$\mu \left(\bigcup_{n=1}^{\infty} A_n \Delta \bigcup_{n=1}^{\infty} C_n \right) \leq \sum_{n=1}^{\infty} \mu(A_n \Delta C_n) < \epsilon/2.$$

Since \mathcal{A} is closed under finite unions we do not know at this point if $\bigcup_{n=1}^{\infty} C_n$ is an element of \mathcal{A} . To solve this problem, we proceed as follows. First note that $\bigcap_{n=1}^m C_n^c \searrow \bigcap_{n=1}^{\infty} C_n^c$, hence

$$\mu \left(\bigcup_{n=1}^{\infty} A_n \cap \bigcap_{n=1}^{\infty} C_n^c \right) = \lim_{m \rightarrow \infty} \mu \left(\bigcup_{n=1}^{\infty} A_n \cap \bigcap_{n=1}^m C_n^c \right),$$

and therefore,

$$\mu \left(\bigcup_{n=1}^{\infty} A_n \Delta \bigcup_{n=1}^{\infty} C_n \right) = \lim_{m \rightarrow \infty} \mu \left(\left(\bigcup_{n=1}^{\infty} A_n \cap \bigcap_{n=1}^m C_n^c \right) \cup \left(\bigcap_{n=1}^{\infty} A_n^c \cap \bigcup_{n=1}^{\infty} C_n \right) \right).$$

Hence there exists m sufficiently large so that

$$\mu \left(\left(\bigcup_{n=1}^{\infty} A_n \cap \bigcap_{n=1}^m C_n^c \right) \cup \left(\bigcap_{n=1}^{\infty} A_n^c \cap \bigcup_{n=1}^{\infty} C_n \right) \right) < \mu \left(\bigcup_{n=1}^{\infty} A_n \Delta \bigcup_{n=1}^{\infty} C_n \right) + \epsilon/2.$$

Since $\bigcap_{n=1}^{\infty} A_n^c \cap \bigcup_{n=1}^m C_n \subseteq \bigcap_{n=1}^{\infty} A_n^c \cap \bigcup_{n=1}^{\infty} C_n$, we get

$$\mu \left(\left(\bigcup_{n=1}^{\infty} A_n \cap \bigcap_{n=1}^m C_n^c \right) \cup \left(\bigcap_{n=1}^{\infty} A_n^c \cap \bigcup_{n=1}^m C_n \right) \right) < \mu \left(\bigcup_{n=1}^{\infty} A_n \Delta \bigcup_{n=1}^{\infty} C_n \right) + \epsilon/2.$$

Thus,

$$\mu \left(\left(\bigcup_{n=1}^{\infty} A_n \Delta \bigcup_{n=1}^m C_n \right) \right) < \varepsilon,$$

and $\bigcup_{n=1}^m C_n \in \mathcal{A}$ since \mathcal{A} is closed under finite unions. This shows that $\bigcup_{n=1}^{\infty} A_n \in \mathcal{D}$. Thus, \mathcal{D} is a σ -algebra.

By definition of \mathcal{D} we have $\mathcal{D} \subseteq \mathcal{B}$. Since $\mathcal{A} \subseteq \mathcal{D}$, and \mathcal{B} is the smallest σ -algebra containing \mathcal{A} we have $\mathcal{B} \subseteq \mathcal{D}$. Therefore, $\mathcal{B} = \mathcal{D}$. \square

1.3 Basic Examples

Example 1.3.1 (Translations) Let $X = [0, 1)$ with the Lebesgue σ -algebra \mathcal{B} , and Lebesgue measure λ . Let $0 < \theta < 1$, define $T : X \rightarrow X$ by

$$Tx = x + \theta \bmod 1 = x + \theta - [x + \theta].$$

Then, by considering intervals it is easy to see that T is measurable and measure preserving.

Example 1.3.2 (Multiplication by 2 modulo 1) Let $(X, \mathcal{B}, \lambda)$ be as in Example (a), and let $T : X \rightarrow X$ be given by

$$Tx = 2x \bmod 1 = \begin{cases} 2x & 0 \leq x < 1/2 \\ 2x - 1 & 1/2 \leq x < 1. \end{cases}$$

For any interval $[a, b)$,

$$T^{-1}[a, b) = \left[\frac{a}{2}, \frac{b}{2} \right) \cup \left[\frac{a+1}{2}, \frac{b+1}{2} \right),$$

and

$$\lambda(T^{-1}[a, b)) = b - a = \lambda([a, b)).$$

Although this map is very simple, it has in fact many facets. For example, iterations of this map yield the binary expansion of points in $[0, 1)$ i.e., using T one can associate with each point in $[0, 1)$ an infinite sequence of 0's and 1's. To do so, we define the function a_1 by

$$a_1(x) = \begin{cases} 0 & \text{if } 0 \leq x < 1/2 \\ 1 & \text{if } 1/2 \leq x < 1, \end{cases}$$

then $Tx = 2x - a_1(x)$. Now, for $n \geq 1$ set $a_n(x) = a_1(T^{n-1}x)$. Fix $x \in X$, for simplicity, we write a_n instead of $a_n(x)$, then $Tx = 2x - a_1$. Rewriting we get $x = \frac{a_1}{2} + \frac{Tx}{2}$. Similarly, $Tx = \frac{a_2}{2} + \frac{T^2x}{2}$. Continuing in this manner, we see that for each $n \geq 1$,

$$x = \frac{a_1}{2} + \frac{a_2}{2^2} + \cdots + \frac{a_n}{2^n} + \frac{T^n x}{2^n}.$$

Since $0 < T^n x < 1$, we get

$$x - \sum_{i=1}^n \frac{a_i}{2^i} = \frac{T^n x}{2^n} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

Thus, $x = \sum_{i=1}^{\infty} \frac{a_i}{2^i}$. We shall later see that the sequence of digits a_1, a_2, \dots forms an i.i.d. sequence of Bernoulli random variables.

Exercise 1.3.1 (Baker's Transformation) Consider the probability space $[0, 1)^2, \mathcal{B} \times \mathcal{B}, \lambda \times \lambda$, where $\mathcal{B} \times \mathcal{B}$ is the product Borel σ -algebra and $\lambda \times \lambda$ the product Lebesgue measure. Define $T : [0, 1)^2 \rightarrow [0, 1)^2$ by

$$T(x, y) = \begin{cases} (2x, \frac{y}{2}) & 0 \leq x < 1/2 \\ (2x - 1, \frac{y+1}{2}) & 1/2 \leq x < 1. \end{cases}$$

Show that T is invertible, measurable and measure preserving.

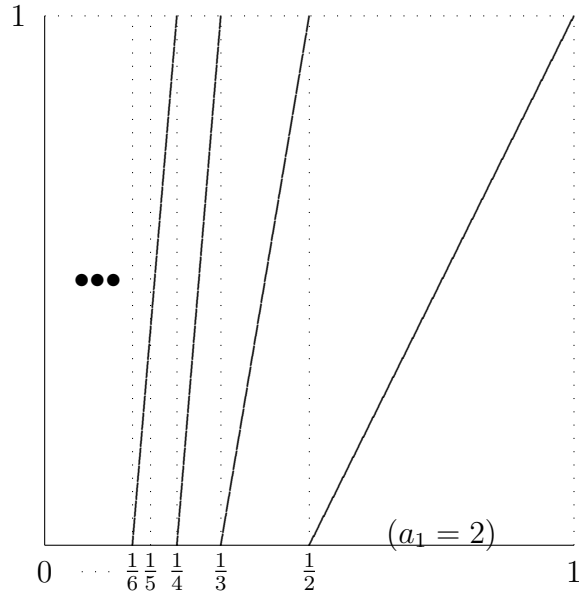
Example 1.3.3 (Lüroth Series) Another kind of series expansion, introduced by J. Lüroth [L] in 1883, motivates this approach. Several authors have studied the dynamics of such systems. Take as partition of $[0, 1)$ the intervals $[\frac{1}{n+1}, \frac{1}{n})$ where $n \in \mathbb{N}$. Every number $x \in [0, 1)$ can be written as a finite or infinite series, the so-called Lüroth (series) expansion

$$x = \frac{1}{a_1(x)} + \frac{1}{a_1(x)(a_1(x) - 1)a_2(x)} + \cdots \\ + \frac{1}{a_1(x)(a_1(x) - 1) \cdots a_{n-1}(x)(a_{n-1}(x) - 1)a_n(x)} + \cdots;$$

here $a_k(x) \geq 2$ for each $k \geq 1$. How is such a series generated?

Let $T : [0, 1) \rightarrow [0, 1)$ be defined by

$$Tx = \begin{cases} n(n+1)x - n, & x \in [\frac{1}{n+1}, \frac{1}{n}), \\ 0, & x = 0. \end{cases} \quad (1.1)$$

Figure 1.1: The Lüroth Series map T .

Let $x \neq 0$, for $k \geq 1$ and $T^{k-1}x \neq 0$ we define the digits $a_n = a_n(x)$ by

$$a_k(x) = a_1(T^{k-1}x),$$

where $a_1(x) = n$ if $x \in [\frac{1}{n}, \frac{1}{n-1})$, $n \geq 2$. Now (1.1) can be written as

$$Tx = \begin{cases} a_1(x)(a_1(x) - 1)x - (a_1(x) - 1), & x \neq 0, \\ 0, & x = 0. \end{cases}$$

Thus¹, for any $x \in (0, 1)$ such that $T^{k-1}x \neq 0$, we have

$$\begin{aligned} x &= \frac{1}{a_1} + \frac{Tx}{a_1(a_1-1)} = \frac{1}{a_1} + \frac{1}{a_1(a_1-1)} \left(\frac{1}{a_2} + \frac{T^2x}{a_2(a_2-1)} \right) \\ &= \frac{1}{a_1} + \frac{1}{a_1(a_1-1)a_2} + \frac{T^2x}{a_1(a_1-1)a_2(a_2-1)} \\ &\quad \vdots \\ &= \frac{1}{a_1} + \cdots + \frac{1}{a_1(a_1-1) \cdots a_{k-1}(a_{k-1}-1)a_k} \\ &\quad + \frac{T^kx}{a_1(a_1-1) \cdots a_k(a_k-1)}. \end{aligned}$$

Notice that, if $T^{k-1}x = 0$ for some $k \geq 1$, and if we assume that k is the smallest positive integer with this property, then

$$x = \frac{1}{a_1} + \cdots + \frac{1}{a_1(a_1-1) \cdots a_{k-1}(a_{k-1}-1)a_k}.$$

In case $T^{k-1}x \neq 0$ for all $k \geq 1$, one gets

$$x = \frac{1}{a_1} + \frac{1}{a_1(a_1-1)a_2} + \cdots + \frac{1}{a_1(a_1-1) \cdots a_{k-1}(a_{k-1}-1)a_k} + \cdots,$$

where $a_k \geq 2$ for each $k \geq 1$. The above infinite series indeed converges to x . To see this, let $S_k = S_k(x)$ be the sum of the first k terms of the sum. Then

$$|x - S_k| = \left| \frac{T^kx}{a_1(a_1-1) \cdots a_k(a_k-1)} \right|;$$

since $T^kx \in [0, 1)$ and $a_k \geq 2$ for all x and all $k \geq 1$, we find

$$|x - S_k| \leq \frac{1}{2^k} \rightarrow 0 \text{ as } k \rightarrow \infty.$$

From the above we also see that if x and y have the same Lüroth expansion, then, for each $k \geq 1$,

$$|x - y| \leq \frac{1}{2^{k-1}}$$

and it follows that x equals y .

¹For ease of notation we drop the argument x from the functions $a_k(x)$.

Exercise 1.3.2 Show that the map T given in Example 1.3.3 is measure preserving with respect to Lebesgue measure λ .

Example 1.3.4 (β -transformation) Let $X = [0, 1)$ with the Lebesgue σ -algebra \mathcal{B} . Let $\beta = \frac{1+\sqrt{5}}{2}$, the golden mean. Notice that $\beta^2 = \beta + 1$. Define a transformation $T : X \rightarrow X$ by

$$Tx = \beta x \bmod 1 = \begin{cases} \beta x & 0 \leq x < 1/\beta \\ \beta x - 1 & 1/\beta \leq x < 1. \end{cases}$$

Then, T is **not** measure preserving with respect to Lebesgue measure (give a counterexample), but is measure preserving with respect to the measure μ given by

$$\mu(B) = \int_B g(x) \, dx,$$

where

$$g(x) = \begin{cases} \frac{5+3\sqrt{5}}{10} & 0 \leq x < 1/\beta \\ \frac{5+\sqrt{5}}{10} & 1/\beta \leq x < 1. \end{cases}$$

Exercise 1.3.3 Verify that T is measure preserving with respect to μ , and show that (similar to Example 1.3.2) iterations of this map generate expansions for points $x \in [0, 1)$ (known as β -expansions) of the form

$$x = \sum_{i=1}^{\infty} \frac{b_i}{\beta^i},$$

where $b_i \in \{0, 1\}$ and $b_i b_{i+1} = 0$ for all $i \geq 1$.

Example 1.3.5 (Continued Fractions) Consider $([0, 1), \mathcal{B})$, where \mathcal{B} is the Lebesgue σ -algebra. Define a transformation $T : [0, 1) \rightarrow [0, 1)$ by $T0 = 0$ and for $x \neq 0$

$$Tx = \frac{1}{x} - \lfloor \frac{1}{x} \rfloor.$$

Exercise 1.3.4 Show that T is **not** measure preserving with respect to Lebesgue measure, but is measure preserving with respect to the so called Gauss probability measure μ given by

$$\mu(B) = \int_B \frac{1}{\log 2} \frac{1}{1+x} \, dx.$$

An interesting feature of this map is that its iterations generate the continued fraction expansion for points in $(0, 1)$. For if we define

$$a_1 = a_1(x) = \begin{cases} 1 & \text{if } x \in (\frac{1}{2}, 1) \\ n & \text{if } x \in (\frac{1}{n+1}, \frac{1}{n}], n \geq 2, \end{cases}$$

then, $Tx = \frac{1}{x} - a_1$ and hence $x = \frac{1}{a_1 + Tx}$. For $n \geq 1$, let $a_n = a_n(x) = a_1(T^{n-1}x)$. Then, after n iterations we see that

$$x = \frac{1}{a_1 + Tx} = \dots = \frac{1}{a_1 + \frac{1}{a_2 + \dots + \frac{1}{a_n + T^n x}}}$$

In fact, in Chapter 4 we will show that if $\frac{p_n}{q_n} = \frac{1}{a_1 + \frac{1}{a_2 + \dots + \frac{1}{a_n}}}$, then the

sequence $\{q_n\}$ is monotonically increasing, and

$$\left| x - \frac{p_n}{q_n} \right| < \frac{1}{q_n^2} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

The last statement implies that

$$x = \frac{1}{a_1 + \frac{1}{a_2 + \frac{1}{a_3 + \dots}}}$$

Example 1.3.6 (Bernoulli Shifts) Let $X = \{0, 1, \dots, k-1\}^{\mathbb{Z}}$ (or $X = \{0, 1, \dots, k-1\}^{\mathbb{N}}$), \mathcal{F} the σ -algebra generated by the cylinders. Let $p = (p_0, p_1, \dots, p_{k-1})$ be a positive probability vector, define a measure μ on \mathcal{F} by specifying it on the cylinder sets as follows

$$\mu(\{x : x_{-n} = a_{-n}, \dots, x_n = a_n\}) = p_{a_{-n}} \dots p_{a_n}.$$

Let $T : X \rightarrow X$ be defined by $Tx = y$, where $y_n = x_{n+1}$. The map T , called the *left shift*, is measurable and measure preserving, since

$$T^{-1}\{x : x_{-n} = a_{-n}, \dots, x_n = a_n\} = \{x : x_{-n+1} = a_{-n}, \dots, x_{n+1} = a_n\},$$

and

$$\mu(\{x : x_{-n+1} = a_{-n}, \dots, x_{n+1} = a_n\}) = p_{a_{-n}} \cdots p_{a_n}.$$

Notice that in case $X = \{0, 1, \dots, k-1\}^{\mathbb{N}}$, then one should consider cylinder sets of the form $\{x : x_0 = a_0, \dots, x_n = a_n\}$. In this case

$$T^{-1}\{x : x_0 = a_0, \dots, x_n = a_n\} = \bigcup_{j=0}^{k-1} \{x : x_0 = j, x_1 = a_0, \dots, x_{n+1} = a_n\},$$

and it is easy to see that T is measurable and measure preserving.

Example 1.3.7 (Markov Shifts) Let (X, \mathcal{F}, T) be as in example (e). We define a measure ν on \mathcal{F} as follows. Let $P = (p_{ij})$ be a stochastic $k \times k$ matrix, and $q = (q_0, q_1, \dots, q_{k-1})$ a positive probability vector such that $qP = q$. Define ν on cylinders by

$$\nu(\{x : x_{-n} = a_{-n}, \dots, x_n = a_n\}) = q_{a_{-n}} p_{a_{-n}a_{-n+1}} \cdots p_{a_{n-1}a_n}.$$

Just as in example 1.3.6, one sees that T is measurable and measure preserving.

1.4 Recurrence

Let T be a measure preserving transformation on a probability space (X, \mathcal{F}, μ) , and let $B \in \mathcal{F}$. A point $x \in B$ is said to be *B-recurrent* if there exists $k \geq 1$ such that $T^k x \in B$.

Theorem 1.4.1 (Poincaré Recurrence Theorem) *If $\mu(B) > 0$, then a.e. $x \in B$ is B-recurrent.*

Proof. Let F be the subset of B consisting of all elements that are not B -recurrent. Then,

$$F = \{x \in B : T^k x \notin B \text{ for all } k \geq 1\}.$$

We want to show that $\mu(F) = 0$. First notice that $F \cap T^{-k}F = \emptyset$ for all $k \geq 1$, hence $T^{-l}F \cap T^{-m}F = \emptyset$ for all $l \neq m$. Thus, the sets $F, T^{-1}F, \dots$ are pairwise disjoint, and $\mu(T^{-n}F) = \mu(F)$ for all $n \geq 1$ (T is measure preserving). If $\mu(F) > 0$, then

$$1 = \mu(X) \geq \mu\left(\bigcup_{k \geq 0} T^{-k}F\right) = \sum_{k \geq 0} \mu(F) = \infty,$$

a contradiction. \square

The proof of the above theorem implies that almost every $x \in B$ returns to B infinitely often. In other words, there exist infinitely many integers $n_1 < n_2 < \dots$ such that $T^{n_i}x \in B$. To see this, let

$$D = \{x \in B : T^k x \in B \text{ for finitely many } k \geq 1\}.$$

Then,

$$D = \{x \in B : T^k x \in F \text{ for some } k \geq 0\} \subseteq \cup_{k=0}^{\infty} T^{-k} F.$$

Thus, $\mu(D) = 0$ since $\mu(F) = 0$ and T is measure preserving.

1.5 Induced Transformations

Let T be a measure preserving transformation on the probability space (X, \mathcal{F}, μ) . Let $A \subset X$ with $\mu(A) > 0$. By Poincaré's Recurrence Theorem almost every $x \in A$ returns to A infinitely often under the action of T . For $x \in A$, let $n(x) := \inf\{n \geq 1 : T^n x \in A\}$. We call $n(x)$ the *first return time* of x to A .

Exercise 1.5.1 Show that n is measurable with respect to the σ -algebra $\mathcal{F} \cap A$ on A .

By Poincaré Theorem, $n(x)$ is finite a.e. on A . In the sequel we remove from A the set of measure zero on which $n(x) = \infty$, and we denote the new set again by A . Consider the σ -algebra $\mathcal{F} \cap A$ on A , which is the restriction of \mathcal{F} to A . Furthermore, let μ_A be the probability measure on A , defined by

$$\mu_A(B) = \frac{\mu(B)}{\mu(A)}, \quad \text{for } B \in \mathcal{F} \cap A,$$

so that $(A, \mathcal{F} \cap A, \mu_A)$ is a probability space. Finally, define the induced map $T_A : A \rightarrow A$ by

$$T_A x = T^{n(x)} x, \quad \text{for } x \in A.$$

From the above we see that T_A is defined on A . What kind of a transformation is T_A ?

Exercise 1.5.2 Show that T_A is measurable with respect to the σ -algebra $\mathcal{F} \cap A$.

Proposition 1.5.1 T_A is measure preserving with respect to μ_A .

Proof. For $k \geq 1$, let

$$A_k = \{x \in A : n(x) = k\}$$

$$B_k = \{x \in X \setminus A : Tx, \dots, T^{k-1}x \notin A, T^kx \in A\}.$$

Notice that $A = \bigcup_{k=1}^{\infty} A_k$, and

$$T^{-1}A = A_1 \cup B_1 \quad \text{and} \quad T^{-1}B_n = A_{n+1} \cup B_{n+1}. \quad (1.2)$$

Let $C \in \mathcal{F} \cap A$, since T is measure preserving it follows that $\mu(C) = \mu(T^{-1}C)$.

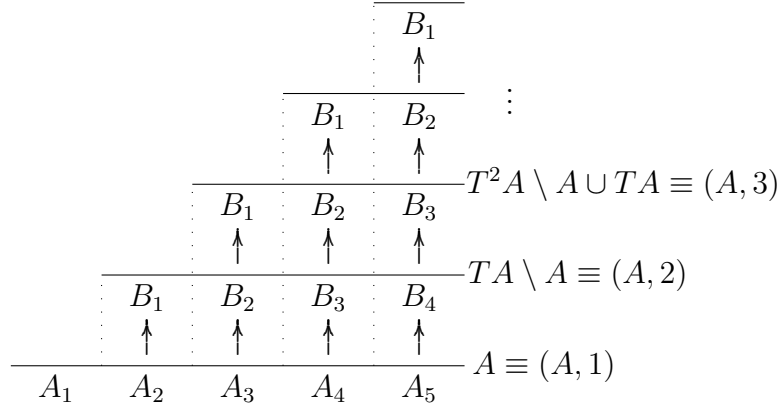


Figure 1.2: A tower.

To show that $\mu_A(C) = \mu_A(T_A^{-1}C)$, we show that

$$\mu(T_A^{-1}C) = \mu(T^{-1}C).$$

Now,

$$T_A^{-1}(C) = \bigcup_{k=1}^{\infty} A_k \cap T_A^{-1}C = \bigcup_{k=1}^{\infty} A_k \cap T^{-k}C,$$

hence

$$\mu(T_A^{-1}(C)) = \sum_{k=1}^{\infty} \mu(A_k \cap T^{-k}C).$$

On the other hand, using repeatedly (1.2), one gets for any $n \geq 1$,

$$\begin{aligned} \mu(T^{-1}(C)) &= \mu(A_1 \cap T^{-1}C) + \mu(B_1 \cap T^{-1}C) \\ &= \mu(A_1 \cap T^{-1}C) + \mu(T^{-1}(B_1 \cap T^{-1}C)) \\ &= \mu(A_1 \cap T^{-1}C) + \mu(A_2 \cap T^{-2}C) + \mu(B_2 \cap T^{-2}C) \\ &\quad \vdots \\ &= \sum_{k=1}^n \mu(A_k \cap T^{-k}C) + \mu(B_n \cap T^{-n}C). \end{aligned}$$

Since

$$1 \geq \mu\left(\bigcup_{n=1}^{\infty} B_n \cap T^{-n}C\right) = \sum_{n=1}^{\infty} \mu(B_n \cap T^{-n}C),$$

it follows that

$$\lim_{n \rightarrow \infty} \mu(B_n \cap T^{-n}C) = 0.$$

Thus,

$$\mu(C) = \mu(T^{-1}C) = \sum_{k=1}^{\infty} \mu(A_k \cap T^{-k}C) = \mu(T_A^{-1}C).$$

This shows that $\mu_A(C) = \mu_A(T_A^{-1}C)$, which implies that T_A is measure preserving with respect to μ_A . \square

Exercise 1.5.3 Assume T is invertible. Without using Proposition 1.5.1 show that for all $C \in \mathcal{F} \cap A$,

$$\mu_A(C) = \mu_A(T_A C).$$

Exercise 1.5.4 Let $G = \frac{1 + \sqrt{5}}{2}$, so that $G^2 = G + 1$. Consider the set

$$X = [0, \frac{1}{G}) \times [0, 1) \cup [\frac{1}{G}, 1) \times [0, \frac{1}{G}),$$

endowed with the product Borel σ -algebra, and the normalized Lebesgue measure $\lambda \times \lambda$. Define the transformation

$$\mathcal{T}(x, y) = \begin{cases} (Gx, \frac{y}{G}), & (x, y) \in [0, \frac{1}{G}) \times [0, 1) \\ (Gx - 1, \frac{1+y}{G}), & (x, y) \in [\frac{1}{G}, 1) \times [0, \frac{1}{G}). \end{cases}$$

- (a) Show that \mathcal{T} is measure preserving with respect to $\lambda \times \lambda$.
- (b) Determine explicitly the induced transformation of \mathcal{T} on the set $[0, 1) \times [0, \frac{1}{G})$.

Exercise 1.5.5 Let $\theta \in (0, 1)$ be irrational. Consider the probability space $([0, 1), \mathcal{B}, \lambda)$, where \mathcal{B} is the Borel σ -algebra and λ is Lebesgue measure restricted to $[0, 1)$. Let $T : [0, 1) \rightarrow [0, 1)$ be translation by $\theta \in (0, 1)$, i.e. $Tx = x + \theta \bmod 1$. Determine explicitly the induced transformation T_A of T on the interval $A = [0, \theta)$.

1.6 Ergodicity

Definition 1.6.1 Let T be a measure preserving transformation on a probability space (X, \mathcal{F}, μ) . The map T is said to be ergodic if for every measurable set A satisfying $T^{-1}A = A$, we have $\mu(A) = 0$ or 1 .

Theorem 1.6.1 Let (X, \mathcal{F}, μ) be a probability space and $T : X \rightarrow X$ measure preserving. The following are equivalent:

- (i) T is ergodic.
- (ii) If $B \in \mathcal{F}$ with $\mu(T^{-1}B \Delta B) = 0$, then $\mu(B) = 0$ or 1 .
- (iii) If $A \in \mathcal{F}$ with $\mu(A) > 0$, then $\mu(\cup_{n=1}^{\infty} T^{-n}A) = 1$.
- (iv) If $A, B \in \mathcal{F}$ with $\mu(A) > 0$ and $\mu(B) > 0$, then there exists $n > 0$ such that $\mu(T^{-n}A \cap B) > 0$.

Remark 1.6.1

1. In case T is invertible, then in the above characterization one can replace T^{-n} by T^n .
2. Note that if $\mu(B \Delta T^{-1}B) = 0$, then $\mu(B \setminus T^{-1}B) = \mu(T^{-1}B \setminus B) = 0$. Since

$$B = (B \setminus T^{-1}B) \cup (B \cap T^{-1}B),$$

and

$$T^{-1}B = (T^{-1}B \setminus B) \cup (B \cap T^{-1}B),$$

we see that after removing a set of measure 0 from B and a set of measure 0 from $T^{-1}B$, the remaining parts are equal. In this case we say that B equals $T^{-1}B$ modulo sets of measure 0.

3. In words, (iii) says that if A is a set of positive measure, almost every $x \in X$ eventually (in fact infinitely often) will visit A .
4. (iv) says that elements of B will eventually enter A .

Proof of Theorem 1.6.1

(i) \Rightarrow (ii) Let $B \in \mathcal{F}$ be such that $\mu(B \Delta T^{-1}B) = 0$. We shall define a measurable set C with $C = T^{-1}C$ and $\mu(C \Delta B) = 0$. Let

$$C = \{x \in X : T^n x \in B \text{ i.o.}\} = \bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} T^{-k}B.$$

Then, $T^{-1}C = C$, hence by (i) $\mu(C) = 0$ or 1. Furthermore,

$$\begin{aligned} \mu(C \Delta B) &= \mu\left(\bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} T^{-k}B \cap B^c\right) + \mu\left(\bigcup_{n=1}^{\infty} \bigcap_{k=n}^{\infty} T^{-k}B^c \cap B\right) \\ &\leq \mu\left(\bigcup_{k=1}^{\infty} T^{-k}B \cap B^c\right) + \mu\left(\bigcup_{k=1}^{\infty} T^{-k}B^c \cap B\right) \\ &\leq \sum_{k=1}^{\infty} \mu(T^{-k}B \Delta B). \end{aligned}$$

Using induction (and the fact that $\mu(E \Delta F) \leq \mu(E \Delta G) + \mu(G \Delta F)$), one can show that for each $k \geq 1$ one has $\mu(T^{-k}B \Delta B) = 0$. Hence, $\mu(C \Delta B) = 0$ which implies that $\mu(C) = \mu(B)$. Therefore, $\mu(B) = 0$ or 1.

(ii) \Rightarrow (iii) Let $\mu(A) > 0$ and let $B = \bigcup_{n=1}^{\infty} T^{-n}A$. Then $T^{-1}B \subset B$. Since T is measure preserving, then $\mu(B) > 0$ and

$$\mu(T^{-1}B \Delta B) = \mu(B \setminus T^{-1}B) = \mu(B) - \mu(T^{-1}B) = 0.$$

Thus, by (ii) $\mu(B) = 1$.

(iii) \Rightarrow (iv) Suppose $\mu(A)\mu(B) > 0$. By (iii)

$$\mu(B) = \mu\left(B \cap \bigcup_{n=1}^{\infty} T^{-n}A\right) = \mu\left(\bigcup_{n=1}^{\infty} (B \cap T^{-n}A)\right) > 0.$$

Hence, there exists $k \geq 1$ such that $\mu(B \cap T^{-k}A) > 0$.

(iv) \Rightarrow (i) Suppose $T^{-1}A = A$ with $\mu(A) > 0$. If $\mu(A^c) > 0$, then by (iv) there exists $k \geq 1$ such that $\mu(A^c \cap T^{-k}A) > 0$. Since $T^{-k}A = A$, it follows that $\mu(A^c \cap A) > 0$, a contradiction. Hence, $\mu(A) = 1$ and T is ergodic. \square

1.7 Other Characterizations of Ergodicity

We denote by $L^0(X, \mathcal{F}, \mu)$ the space of all complex valued measurable functions on the probability space (X, \mathcal{F}, μ) . Let

$$L^p(X, \mathcal{F}, \mu) = \{f \in L^0(X, \mathcal{F}, \mu) : \int_X |f|^p d\mu(x) < \infty\}.$$

We use the subscript \mathbb{R} whenever we are dealing only with real-valued functions.

Let $(X_i, \mathcal{F}_i, \mu_i)$, $i = 1, 2$ be two probability spaces, and $T : X_1 \rightarrow X_2$ a measure preserving transformation i.e., $\mu_2(A) = \mu_1(T^{-1}A)$. Define the *induced* operator $U_T : L^0(X_2, \mathcal{F}_2, \mu_2) \rightarrow L^0(X_1, \mathcal{F}_1, \mu_1)$ by

$$U_T f = f \circ T.$$

The following properties of U_T are easy to prove.

Proposition 1.7.1 *The operator U_T has the following properties:*

- (i) U_T is linear
- (ii) $U_T(fg) = U_T(f)U_T(g)$
- (iii) $U_T c = c$ for any constant c .
- (iv) U_T is a positive linear operator
- (v) $U_T 1_B = 1_B \circ T = 1_{T^{-1}B}$ for all $B \in \mathcal{F}_2$.
- (vi) $\int_{X_1} U_T f d\mu_1 = \int_{X_2} f d\mu_2$ for all $f \in L^0(X_2, \mathcal{F}_2, \mu_2)$, (where if one side doesn't exist or is infinite, then the other side has the same property).
- (vii) Let $p \geq 1$. Then, $U_T L^p(X_2, \mathcal{F}_2, \mu_2) \subset L^p(X_1, \mathcal{F}_1, \mu_1)$, and $\|U_T f\|_p = \|f\|_p$ for all $f \in L^p(X_2, \mathcal{F}_2, \mu_2)$.

Exercise 1.7.1 Prove Proposition 1.7.1.

Exercise 1.7.2 Let (X, \mathcal{F}, μ) be a probability space, and $T : X \rightarrow X$ a measure preserving transformation. Let $f \in L^1(X, \mathcal{F}, \mu)$. Show that if $f(Tx) \leq f(x)$ μ a.e., then $f(x) = f(Tx)$ μ a.e.

Exercise 1.7.3 Let (X, \mathcal{F}, μ) be a probability space, and $T : X \rightarrow X$ a measure preserving and ergodic transformation. Suppose $f \geq 0$ is a measurable function. Show that if the set $A = \{x \in X : f(x) > 0\}$ has positive μ measure, then for μ a.e. x , one has

$$\sum_{n=1}^{\infty} f(T^n x) = \infty.$$

In the following theorem, we give a new characterization of ergodicity

Theorem 1.7.1 *Let (X, \mathcal{F}, μ) be a probability space, and $T : X \rightarrow X$ measure preserving. The following are equivalent:*

- (i) T is ergodic.
- (ii) If $f \in L^0(X, \mathcal{F}, \mu)$, with $f(Tx) = f(x)$ for all x , then f is a constant a.e.
- (iii) If $f \in L^0(X, \mathcal{F}, \mu)$, with $f(Tx) = f(x)$ for a.e. x , then f is a constant a.e.
- (iv) If $f \in L^2(X, \mathcal{F}, \mu)$, with $f(Tx) = f(x)$ for all x , then f is a constant a.e.
- (v) If $f \in L^2(X, \mathcal{F}, \mu)$, with $f(Tx) = f(x)$ for a.e. x , then f is a constant a.e.

Proof. The implications (iii) \Rightarrow (ii), (ii) \Rightarrow (iv), (v) \Rightarrow (iv), and (iii) \Rightarrow (v) are all clear. It remains to show (i) \Rightarrow (iii) and (iv) \Rightarrow (i).

(i) \Rightarrow (iii) Suppose $f(Tx) = f(x)$ a.e. and assume without any loss of generality that f is real (otherwise we consider separately the real and imaginary parts of f). For each $n \geq 1$ and $k \in \mathbb{Z}$, let

$$I_{k,n} = \left\{ x \in X : \frac{k}{2^n} \leq f(x) < \frac{k+1}{2^n} \right\}.$$

Then, $T^{-1}I_{k,n} \Delta I_{k,n} \subseteq \{x : f(Tx) \neq f(x)\}$ which implies that

$$\mu(T^{-1}I_{k,n} \Delta I_{k,n}) = 0.$$

By ergodicity of T , $\mu(I_{k,n}) = 0$ or 1 , for each $k \in \mathbb{Z}$. On the other hand, for each $n \geq 1$, we have

$$X = \bigcup_{k \in \mathbb{Z}} I_{k,n} \text{ (disjoint union).}$$

Hence, for each $n \geq 1$, there exists a unique integer k_n such that $\mu(I_{k_n,n}) = 1$.

In fact, $I_{k_1,1} \supseteq I_{k_2,2} \supseteq \dots$, and $\{\frac{k_n}{2^n}\}$ is a bounded increasing sequence, hence $\lim_{n \rightarrow \infty} k_n/2^n$ exists. Let $Y = \bigcap_{n \geq 1} I_{k_n,n}$, then $\mu(Y) = 1$. Now, if $x \in Y$, then $0 \leq |f(x) - k_n/2^n| < 1/2^n$ for all n . Hence, $f(x) = \lim_{n \rightarrow \infty} k_n/2^n$, and f is a constant on Y .

(iv) \Rightarrow (i) Suppose $T^{-1}A = A$ and $\mu(A) > 0$. We want to show that $\mu(A) = 1$. Consider 1_A , the indicator function of A . We have $1_A \in L^2(X, \mathcal{F}, \mu)$, and $1_A \circ T = 1_{T^{-1}A} = 1_A$. Hence, by (iv), 1_A is a constant a.e., hence $1_A = 1$ a.e. and therefore $\mu(A) = 1$. \square

1.8 Examples of Ergodic Transformations

Example 1.8.1 (Irrational Rotations) Consider $([0, 1], \mathcal{B}, \lambda)$, where \mathcal{B} is the Lebesgue σ -algebra, and λ Lebesgue measure. For $\theta \in (0, 1)$, consider the transformation $T_\theta : [0, 1] \rightarrow [0, 1]$ defined by $T_\theta x = x + \theta \pmod{1}$. We have seen in Example 1.3.1 that T_θ is measure preserving with respect λ . When is T_θ ergodic?

If θ is rational, then T_θ is not ergodic. Consider for example $\theta = 1/4$, then the set

$$A = [0, 1/8) \cup [1/4, 3/8) \cup [1/2, 5/8) \cup [3/4, 7/8)$$

is T_θ -invariant but $\mu(A) = 1/2$.

Exercise 1.8.1 Suppose $\theta = p/q$ with $\gcd(p, q) = 1$. Find a non-trivial T_θ -invariant set. Conclude that T_θ is not ergodic if θ is a rational.

Claim. T_θ is ergodic if and only if θ is irrational.

Proof of Claim.

(\Rightarrow) The contrapositive statement is given in Exercise 1.8.1 i.e., if θ is rational, then T_θ is not ergodic.

(\Leftarrow) Suppose θ is irrational, and let $f \in L^2(X, \mathcal{B}, \lambda)$ be T_θ -invariant. Write f in its Fourier series

$$f(x) = \sum_{n \in \mathbb{Z}} a_n e^{2\pi i n x}.$$

Since $f(T_\theta x) = f(x)$, then

$$\begin{aligned} f(T_\theta x) &= \sum_{n \in \mathbb{Z}} a_n e^{2\pi i n(x+\theta)} = \sum_{n \in \mathbb{Z}} a_n e^{2\pi i n \theta} e^{2\pi i n x} \\ &= f(x) = \sum_{n \in \mathbb{Z}} a_n e^{2\pi i n x}. \end{aligned}$$

Hence, $\sum_{n \in \mathbb{Z}} a_n (1 - e^{2\pi i n \theta}) e^{2\pi i n x} = 0$. By the uniqueness of the Fourier coefficients, we have $a_n (1 - e^{2\pi i n \theta}) = 0$, for all $n \in \mathbb{Z}$. If $n \neq 0$, since θ is irrational we have $1 - e^{2\pi i n \theta} \neq 0$. Thus, $a_n = 0$ for all $n \neq 0$, and therefore $f(x) = a_0$ is a constant. By Theorem 1.7.1, T_θ is ergodic.

Exercise 1.8.2 Consider the probability space $([0, 1], \mathcal{B} \times \mathcal{B}, \lambda \times \lambda)$, where as above \mathcal{B} is the Lebesgue σ -algebra on $[0, 1]$, and λ normalized Lebesgue measure. Suppose $\theta \in (0, 1)$ is irrational, and define $T_\theta \times T_\theta : [0, 1] \times [0, 1] \rightarrow [0, 1] \times [0, 1]$ by

$$T_\theta \times T_\theta(x, y) = (x + \theta \bmod(1), y + \theta \bmod(1)).$$

Show that $T_\theta \times T_\theta$ is measure preserving, but is **not** ergodic.

Exercise 1.8.3 Consider the probability space $([0, 1] \times [0, 1], \mathcal{B} \times \mathcal{B}, \lambda \times \lambda)$, where $\mathcal{B} \times \mathcal{B}$ is the two-dimensional Borel σ -algebra and $\lambda \times \lambda$ is the two-dimensional Lebesgue measure restricted to $[0, 1] \times [0, 1]$. Prove that the transformation $S : [0, 1] \times [0, 1] \rightarrow [0, 1] \times [0, 1]$ given by $S(x, y) = (x + \theta \bmod 1, x + y \bmod 1)$ with θ irrational is measure preserving and ergodic with respect to $\lambda \times \lambda$. (Hint: The Fourier series $\sum_{n,m} c_{n,m} e^{2\pi i(n x + m y)}$ of a function $f \in L^2([0, 1] \times [0, 1], \mathcal{B} \times \mathcal{B}, \lambda \times \lambda)$ satisfies $\sum_{n,m} |c_{n,m}|^2 < \infty$.)

Example 1.8.2 (One (or two) sided shift) Let $X = \{0, 1, \dots, k-1\}^{\mathbb{N}}$, \mathcal{F} the σ -algebra generated by the cylinders, and μ the product measure defined on cylinder sets by

$$\mu(\{x : x_0 = a_0, \dots, x_n = a_n\}) = p_{a_0} \cdots p_{a_n},$$

where $p = (p_0, p_1, \dots, p_{k-1})$ is a positive probability vector. Consider the left shift T defined on X by $Tx = y$, where $y_n = x_{n+1}$ (See Example (e) in Subsection 1.3). We show that T is ergodic. Let E be a measurable subset of X which is T -invariant i.e., $T^{-1}E = E$. For any $\epsilon > 0$, by Lemma 1.2.1 (see subsection 1.2), there exists $A \in \mathcal{F}$ which is a finite disjoint union of cylinders such that $\mu(E\Delta A) < \epsilon$. Then

$$\begin{aligned} |\mu(E) - \mu(A)| &= |\mu(E \setminus A) - \mu(A \setminus E)| \\ &\leq \mu(E \setminus A) + \mu(A \setminus E) = \mu(E\Delta A) < \epsilon. \end{aligned}$$

Since A depends on finitely many coordinates only, there exists $n_0 > 0$ such that $T^{-n_0}A$ depends on different coordinates than A . Since μ is a product measure, we have

$$\mu(A \cap T^{-n_0}A) = \mu(A)\mu(T^{-n_0}A) = \mu(A)^2.$$

Further,

$$\mu(E\Delta T^{-n_0}A) = \mu(T^{-n_0}E\Delta T^{-n_0}A) = \mu(E\Delta A) < \epsilon,$$

and

$$\mu(E\Delta(A \cap T^{-n_0}A)) \leq \mu(E\Delta A) + \mu(E\Delta T^{-n_0}A) < 2\epsilon.$$

Hence,

$$|\mu(E) - \mu((A \cap T^{-n_0}A))| \leq \mu(E\Delta(A \cap T^{-n_0}A)) < 2\epsilon.$$

Thus,

$$\begin{aligned} |\mu(E) - \mu(E)^2| &\leq |\mu(E) - \mu(A)^2| + |\mu(A)^2 - \mu(E)^2| \\ &= |\mu(E) - \mu((A \cap T^{-n_0}A))| + (\mu(A) + \mu(E))|\mu(A) - \mu(E)| \\ &< 4\epsilon. \end{aligned}$$

Since $\epsilon > 0$ is arbitrary, it follows that $\mu(E) = \mu(E)^2$, hence $\mu(E) = 0$ or 1 . Therefore, T is ergodic.

Example 1.8.3 (Induced transformations of ergodic transformations) Let T be an ergodic measure preserving transformation on the probability space (X, \mathcal{F}, μ) , and $A \in \mathcal{F}$ with $\mu(A) > 0$. Consider the induced transformation T_A on $(A, \mathcal{F} \cap A, \mu_A)$ of T (see subsection 1.5). Recall that $T_A x = T^{n(x)}x$, where $n(x) := \inf\{n \geq 1 : T^n x \in A\}$. Let (as in the proof of Proposition 1.5.1),

$$A_k = \{x \in A : n(x) = k\}$$

$$B_k = \{x \in X \setminus A : Tx, \dots, T^{k-1}x \notin A, T^k x \in A\}.$$

Proposition 1.8.1 *If T is ergodic on (X, \mathcal{F}, μ) , then T_A is ergodic on $(A, \mathcal{F} \cap A, \mu_A)$.*

Proof. Let $C \in \mathcal{F} \cap A$ be such that $T_A^{-1}C = C$. We want to show that $\mu_A(C) = 0$ or 1; equivalently, $\mu(C) = 0$ or $\mu(C) = \mu(A)$. Since $A = \bigcup_{k \geq 1} A_k$, we have $C = T_A^{-1}C = \bigcup_{k \geq 1} A_k \cap T^{-k}C$. Let $E = \bigcup_{k \geq 1} B_k \cap T^{-k}C$, and $F = E \cup C$ (disjoint union). Recall that (see subsection 1.5) $T^{-1}A = A_1 \cup B_1$, and $T^{-1}B_k = A_{k+1} \cup B_{k+1}$. Hence,

$$\begin{aligned} T^{-1}F &= T^{-1}E \cup T^{-1}C \\ &= \bigcup_{k \geq 1} [(A_{k+1} \cup B_{k+1}) \cap T^{-(k+1)}C] \cup [(A_1 \cup B_1) \cap T^{-1}C] \\ &= \bigcup_{k \geq 1} (A_k \cap T^{-k}C) \cup \bigcup_{k \geq 1} (B_k \cap T^{-k}C) \\ &= C \cup E = F. \end{aligned}$$

Hence, F is T -invariant, and by ergodicity of T we have $\mu(F) = 0$ or 1. If $\mu(F) = 0$, then $\mu(C) = 0$, and hence $\mu_A(C) = 0$. If $\mu(F) = 1$, then $\mu(X \setminus F) = 0$. Since

$$X \setminus F = (A \setminus C) \cup ((X \setminus A) \setminus E) \supseteq A \setminus C,$$

it follows that

$$\mu(A \setminus C) \leq \mu(X \setminus F) = 0.$$

Since $\mu(A \setminus C) = \mu(A) - \mu(C)$, we have $\mu(A) = \mu(C)$, i.e., $\mu_A(C) = 1$. \square

Exercise 1.8.4 Show that if T_A is ergodic and $\mu\left(\bigcup_{k \geq 1} T^{-k}A\right) = 1$, then, T is ergodic.

The following lemma provides, in some cases, a useful tool to verify that a measure preserving transformation defined on $([0, 1], \mathcal{B}, \mu)$ is ergodic, where \mathcal{B} is the Lebesgue σ -algebra, and μ is a probability measure equivalent to Lebesgue measure λ (i.e., $\mu(A) = 0$ if and only if $\lambda(A) = 0$).

Lemma 1.8.1 (Knopp's Lemma) *If B is a Lebesgue set and \mathcal{C} is a class of subintervals of $[0, 1)$, satisfying*

- (a) *every open subinterval of $[0, 1)$ is at most a countable union of disjoint elements from \mathcal{C} ,*
- (b) *$\forall A \in \mathcal{C}$, $\lambda(A \cap B) \geq \gamma\lambda(A)$, where $\gamma > 0$ is independent of A ,*

then $\lambda(B) = 1$.

Proof. The proof is done by contradiction. Suppose $\lambda(B^c) > 0$. Given $\varepsilon > 0$ there exists by Lemma 1.2.1 a set E_ε that is a finite disjoint union of open intervals such that $\lambda(B^c \Delta E_\varepsilon) < \varepsilon$. Now by conditions (a) and (b) (that is, writing E_ε as a countable union of disjoint elements of \mathcal{C}) one gets that $\lambda(B \cap E_\varepsilon) \geq \gamma\lambda(E_\varepsilon)$.

Also from our choice of E_ε and the fact that

$$\lambda(B^c \Delta E_\varepsilon) \geq \lambda(B \cap E_\varepsilon) \geq \gamma\lambda(E_\varepsilon) \geq \gamma\lambda(B^c \cap E_\varepsilon) > \gamma(\lambda(B^c) - \varepsilon),$$

we have that

$$\gamma(\lambda(B^c) - \varepsilon) < \lambda(B^c \Delta E_\varepsilon) < \varepsilon,$$

implying that $\gamma\lambda(B^c) < \varepsilon + \gamma\varepsilon$. Since $\varepsilon > 0$ is arbitrary, we get a contradiction. \square

Example 1.8.4 (Multiplication by 2 modulo 1) Consider $([0, 1), \mathcal{B}, \lambda)$, and let $T : X \rightarrow X$ be given by

$$Tx = 2x \bmod 1 = \begin{cases} 2x & 0 \leq x < 1/2 \\ 2x - 1 & 1/2 \leq x < 1, \end{cases}$$

We have seen that T is measure preserving. We will use Lemma 1.8.1 to show that T is ergodic. Let \mathcal{C} be the collection of all intervals of the form

$[k/2^n, (k+1)/2^n)$ with $n \geq 1$ and $0 \leq k \leq 2^n - 1$. Notice that the set $\{k/2^n : n \geq 1, 0 \leq k < 2^n - 1\}$ of dyadic rationals is dense in $[0, 1)$, hence each open interval is at most a countable union of disjoint elements of \mathcal{C} . Hence, \mathcal{C} satisfies the first hypothesis of Knopp's Lemma. Now, T^n maps each dyadic interval of the form $[k/2^n, (k+1)/2^n)$ linearly onto $[0, 1)$, (we call such an interval dyadic of order n); in fact, $T^n x = 2^n x \bmod(1)$. Let $B \in \mathcal{B}$ be T -invariant, and assume $\lambda(B) > 0$. Let $A \in \mathcal{C}$, and assume that A is dyadic of order n . Then, $T^n A = [0, 1)$ and

$$\begin{aligned} \lambda(A \cap B) &= \lambda(A \cap T^{-n} B) = \frac{1}{\lambda(A)} \lambda(T^n A \cap B) \\ &= \frac{1}{2^n} \lambda(B) = \lambda(A) \lambda(B). \end{aligned}$$

Thus, the second hypothesis of Knopp's Lemma is satisfied with $\gamma = \lambda(B) > 0$. Hence, $\lambda(B) = 1$. Therefore T is ergodic.

Example 1.8.5 (Lüroth series revisited) Consider the map T of Example 1.3.3. In Exercise 1.3.2 we saw that T is measure preserving with respect to Lebesgue measure λ . We now show that T is ergodic with respect to Lebesgue measure λ using Knopp's Lemma. The collection \mathcal{C} consists in this case of all fundamental intervals of all ranks. A fundamental interval of rank k is a set of the form

$$\begin{aligned} \Delta(i_1, i_2, \dots, i_k) &= \Delta(i_1) \cap T^{-1} \Delta(i_2) \cap \dots \cap \Delta(i_k) \\ &= \{x : a_1(x) = i_1, a_2(x) = i_2, \dots, a_k(x) = i_k\}. \end{aligned}$$

Notice that $\Delta(i_1, i_2, \dots, i_k)$ is an interval with end points

$$\frac{P_k}{Q_k} \quad \text{and} \quad \frac{P_k}{Q_k} + \frac{1}{i_1(i_1 - 1) \cdots i_k(i_k - 1)},$$

where

$$P_k/Q_k = \frac{1}{i_1} + \frac{1}{i_1(i_1 - 1)i_2} + \dots + \frac{1}{i_1(i_1 - 1) \cdots i_{k-1}(i_{k-1} - 1)i_k}.$$

Furthermore, $T^k(\Delta(i_1, i_2, \dots, i_k)) = [0, 1)$, and T^k restricted to $\Delta(i_1, i_2, \dots, i_k)$ has slope

$$i_1(i_1 - 1) \cdots i_{k-1}(i_{k-1} - 1)i_k(i_k - 1) = \frac{1}{\lambda(\Delta(i_1, i_2, \dots, i_k))}.$$

Since $\lim_{k \rightarrow \infty} \text{diam}(\Delta(i_1, i_2, \dots, i_k)) = 0$ for any sequence i_1, i_2, \dots , the collection \mathcal{C} generates the Borel σ -algebra. Now let A be a T -invariant Borel set of positive Lebesgue measure, and let E be any fundamental interval of rank n , then

$$\lambda(A \cap E) = \lambda(T^{-n}A \cap E) = \lambda(E)\lambda(A).$$

By Knopp's Lemma with $\gamma = \lambda(A)$ we get that $\lambda(A) = 1$; i.e. T is ergodic with respect to λ . \square

Exercise 1.8.5

Let λ be the normalized Lebesgue measure on $([0, 1], \mathcal{B})$, where \mathcal{B} is the Lebesgue σ -algebra. Consider the transformation $T : [0, 1] \rightarrow [0, 1]$ given by

$$Tx = \begin{cases} 3x & 0 \leq x < 1/3 \\ \frac{3}{2}x - \frac{1}{2} & 1/3 \leq x < 1. \end{cases}$$

For $x \in [0, 1]$ let

$$s_1(x) = \begin{cases} 3 & 0 \leq x < 1/3 \\ \frac{3}{2} & 1/3 \leq x < 1, \end{cases}$$

$$h_1(x) = \begin{cases} 0 & 0 \leq x < 1/3 \\ \frac{1}{2} & 1/3 \leq x < 1, \end{cases}$$

and

$$a_1(x) = \begin{cases} 0 & 0 \leq x < 1/3 \\ 1 & 1/3 \leq x < 1. \end{cases}$$

Let $s_n = s_n(x) = s_1(T^{n-1}x)$, $h_n = h_n(x) = h_1(T^{n-1}x)$ and $a_n = a_n(x) = a_1(T^{n-1}x)$ for $n \geq 1$.

(a) Show that for any $x \in [0, 1]$ one has $x = \sum_{k=1}^{\infty} \frac{h_k}{s_1 s_2 \cdots s_k}$.

(b) Show that T is measure preserving and ergodic with respect to the measure λ .

(c) Show that for each $n \geq 1$ and any sequence $i_1, i_2, \dots, i_n \in \{0, 1\}$ one has

$$\lambda(\{x \in [0, 1] : a_1(x) = i_1, a_2(x) = i_2, \dots, a_n(x) = i_n\}) = \frac{2^k}{3^n},$$

where $k = \#\{1 \leq j \leq n : i_j = 1\}$.

Exercise 1.8.6 Let $\beta > 1$ be a non-integer, and consider the transformation $T_\beta : [0, 1) \rightarrow [0, 1)$ given by $T_\beta x = \beta x \bmod(1) = \beta x - \lfloor \beta x \rfloor$. Use Lemma 1.8.1 to show that T_β is ergodic with respect to Lebesgue measure λ , i.e. if $T_\beta^{-1}A = A$, then $\lambda(A) = 0$ or 1 .

Chapter 2

Ergodic Theorems

2.1 The Pointwise Ergodic Theorem and its consequences

The Pointwise Ergodic Theorem is also known as Birkhoff's Ergodic Theorem or the Individual Ergodic Theorem (1931). This theorem is in fact a generalization of the Strong Law of Large Numbers (SLLN) which states that for a sequence Y_1, Y_2, \dots of i.i.d. random variables on a probability space (X, \mathcal{F}, μ) , with $E|Y_i| < \infty$; one has

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n Y_i = EY_1 \text{ (a.e.)}.$$

For example consider $X = \{0, 1\}^{\mathbb{N}}$, \mathcal{F} the σ -algebra generated by the cylinder sets, and μ the uniform product measure, i.e.,

$$\mu(\{x : x_1 = a_1, x_2 = a_2, \dots, x_n = a_n\}) = 1/2^n.$$

Suppose one is interested in finding the frequency of the digit 1. More precisely, for a.e. x we would like to find

$$\lim_{n \rightarrow \infty} \frac{1}{n} \#\{1 \leq i \leq n : x_i = 1\}.$$

Using the Strong Law of Large Numbers one can answer this question easily. Define

$$Y_i(x) := \begin{cases} 1, & \text{if } x_i = 1, \\ 0, & \text{otherwise.} \end{cases}$$

Since μ is product measure, it is easy to see that Y_1, Y_2, \dots form an i.i.d. Bernoulli process, and $EY_i = E|Y_i| = 1/2$. Further, $\#\{1 \leq i \leq n : x_i = 1\} = \sum_{i=1}^n Y_i(x)$. Hence, by SLLN one has

$$\lim_{n \rightarrow \infty} \frac{1}{n} \#\{1 \leq i \leq n : x_i = 1\} = \frac{1}{2}.$$

Suppose now we are interested in the frequency of the block 011, i.e., we would like to find

$$\lim_{n \rightarrow \infty} \frac{1}{n} \#\{1 \leq i \leq n : x_i = 0, x_{i+1} = 1, x_{i+2} = 1\}.$$

We can start as above by defining random variables

$$Z_i(x) := \begin{cases} 1, & \text{if } x_i = 0, x_{i+1} = 1, x_{i+2} = 1, \\ 0, & \text{otherwise.} \end{cases}$$

Then,

$$\frac{1}{n} \#\{1 \leq i \leq n : x_i = 0, x_{i+1} = 1, x_{i+2} = 1\} = \frac{1}{n} \sum_{i=1}^n Z_i(x).$$

It is not hard to see that this sequence is stationary but not independent. So one cannot directly apply the strong law of large numbers. Notice that if T is the left shift on X , then $Y_n = Y_1 \circ T^{n-1}$ and $Z_n = Z_1 \circ T^{n-1}$.

In general, suppose (X, \mathcal{F}, μ) is a probability space and $T : X \rightarrow X$ a measure preserving transformation. For $f \in L^1(X, \mathcal{F}, \mu)$, we would like to know under

what conditions does the limit $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x)$ exist a.e. If it does exist

what is its value? This is answered by the Pointwise Ergodic Theorem which was originally proved by G.D. Birkhoff in 1931. Since then, several proofs of this important theorem have been obtained; here we present a recent proof given by T. Kamae and M.S. Keane in [KK].

Theorem 2.1.1 (The Pointwise Ergodic Theorem) *Let (X, \mathcal{F}, μ) be a probability space and $T : X \rightarrow X$ a measure preserving transformation. Then, for any f in $L^1(\mu)$,*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i(x)) = f^*(x)$$

exists a.e., is T -invariant and $\int_X f \, d\mu = \int_X f^* \, d\mu$. If moreover T is ergodic, then f^* is a constant a.e. and $f^* = \int_X f \, d\mu$.

For the proof of the above theorem, we need the following simple lemma.

Lemma 2.1.1 *Let $M > 0$ be an integer, and suppose $\{a_n\}_{n \geq 0}$, $\{b_n\}_{n \geq 0}$ are sequences of non-negative real numbers such that for each $n = 0, 1, 2, \dots$ there exists an integer $1 \leq m \leq M$ with*

$$a_n + \cdots + a_{n+m-1} \geq b_n + \cdots + b_{n+m-1}.$$

Then, for each positive integer $N > M$, one has

$$a_0 + \cdots + a_{N-1} \geq b_0 + \cdots + b_{N-M-1}.$$

Proof of Lemma 2.1.1 Using the hypothesis we recursively find integers $m_0 < m_1 < \cdots < m_k < N$ with the following properties

$$m_0 \leq M, \quad m_{i+1} - m_i \leq M \text{ for } i = 0, \dots, k-1, \text{ and } N - m_k < M,$$

$$a_0 + \cdots + a_{m_0-1} \geq b_0 + \cdots + b_{m_0-1},$$

$$a_{m_0} + \cdots + a_{m_1-1} \geq b_{m_0} + \cdots + b_{m_1-1},$$

$$\vdots$$

$$a_{m_{k-1}} + \cdots + a_{m_k-1} \geq b_{m_{k-1}} + \cdots + b_{m_k-1}.$$

Then,

$$\begin{aligned} a_0 + \cdots + a_{N-1} &\geq a_0 + \cdots + a_{m_k-1} \\ &\geq b_0 + \cdots + b_{m_k-1} \geq b_0 + \cdots + b_{N-M-1}. \end{aligned}$$

□

Proof of Theorem 2.1.1 Assume with no loss of generality that $f \geq 0$ (otherwise we write $f = f^+ - f^-$, and we consider each part separately). Let

$f_n(x) = f(x) + \dots + f(T^{n-1}x)$, $\bar{f}(x) = \limsup_{n \rightarrow \infty} \frac{f_n(x)}{n}$, and $\underline{f}(x) = \liminf_{n \rightarrow \infty} \frac{f_n(x)}{n}$. Then \bar{f} and \underline{f} are T -invariant, since

$$\begin{aligned} \bar{f}(Tx) &= \limsup_{n \rightarrow \infty} \frac{f_n(Tx)}{n} \\ &= \limsup_{n \rightarrow \infty} \left[\frac{f_{n+1}(x)}{n+1} \cdot \frac{n+1}{n} - \frac{f(x)}{n} \right] \\ &= \limsup_{n \rightarrow \infty} \frac{f_{n+1}(x)}{n+1} = \bar{f}(x). \end{aligned}$$

Similarly \underline{f} is T -invariant. Now, to prove that f^* exists, is integrable and T -invariant, it is enough to show that

$$\int_X \underline{f} d\mu \geq \int_X f d\mu \geq \int_X \bar{f} d\mu.$$

For since $\bar{f} - \underline{f} \geq 0$, this would imply that $\bar{f} = \underline{f} = f^*$. a.e.

We first prove that $\int_X \bar{f} d\mu \leq \int_X f d\mu$. Fix any $0 < \epsilon < 1$, and let $L > 0$ be any real number. By definition of \bar{f} , for any $x \in X$, there exists an integer $m > 0$ such that

$$\frac{f_m(x)}{m} \geq \min(\bar{f}(x), L)(1 - \epsilon).$$

Now, for any $\delta > 0$ there exists an integer $M > 0$ such that the set

$$X_0 = \{x \in X : \exists 1 \leq m \leq M \text{ with } f_m(x) \geq m \min(\bar{f}(x), L)(1 - \epsilon)\}$$

has measure at least $1 - \delta$. Define F on X by

$$F(x) = \begin{cases} f(x) & x \in X_0 \\ L & x \notin X_0. \end{cases}$$

Notice that $f \leq F$ (why?). For any $x \in X$, let $a_n = a_n(x) = F(T^n x)$, and $b_n = b_n(x) = \min(\bar{f}(x), L)(1 - \epsilon)$ (so b_n is independent of n). We now show that $\{a_n\}$ and $\{b_n\}$ satisfy the hypothesis of Lemma 2.1.1 with $M > 0$ as above. For any $n = 0, 1, 2, \dots$

— if $T^n x \in X_0$, then there exists $1 \leq m \leq M$ such that

$$\begin{aligned} f_m(T^n x) &\geq m \min(\bar{f}(T^n x), L)(1 - \epsilon) \\ &= m \min(\bar{f}(x), L)(1 - \epsilon) \\ &= b_n + \dots + b_{n+m-1}. \end{aligned}$$

Hence,

$$\begin{aligned} a_n + \dots + a_{n+m-1} &= F(T^n x) + \dots + F(T^{n+m-1} x) \\ &\geq f(T^n x) + \dots + f(T^{n+m-1} x) = f_m(T^n x) \\ &\geq b_n + \dots + b_{n+m-1}. \end{aligned}$$

— If $T^n x \notin X_0$, then take $m = 1$ since

$$a_n = F(T^n x) = L \geq \min(\bar{f}(x), L)(1 - \epsilon) = b_n.$$

Hence by Lemma 2.1.1 for all integers $N > M$ one has

$$F(x) + \dots + F(T^{N-1} x) \geq (N - M) \min(\bar{f}(x), L)(1 - \epsilon).$$

Integrating both sides, and using the fact that T is measure preserving, one gets

$$N \int_X F(x) \, d\mu(x) \geq (N - M) \int_X \min(\bar{f}(x), L)(1 - \epsilon) \, d\mu(x).$$

Since

$$\int_X F(x) \, d\mu(x) = \int_{X_0} f(x) \, d\mu(x) + L\mu(X \setminus X_0),$$

one has

$$\begin{aligned} \int_X f(x) \, d\mu(x) &\geq \int_{X_0} f(x) \, d\mu(x) \\ &= \int_X F(x) \, d\mu(x) - L\mu(X \setminus X_0) \\ &\geq \frac{(N - M)}{N} \int_X \min(\bar{f}(x), L)(1 - \epsilon) \, d\mu(x) - L\delta. \end{aligned}$$

Now letting first $N \rightarrow \infty$, then $\delta \rightarrow 0$, then $\epsilon \rightarrow 0$, and lastly $L \rightarrow \infty$ one gets together with the monotone convergence theorem that \bar{f} is integrable, and

$$\int_X f(x) \, d\mu(x) \geq \int_X \bar{f}(x) \, d\mu(x).$$

We now prove that

$$\int_X f(x) \, d\mu(x) \leq \int_X \underline{f}(x) \, d\mu(x).$$

Fix $\epsilon > 0$, and $\delta_0 > 0$. Since $f \geq 0$, there exists $\delta > 0$ such that whenever $A \in \mathcal{F}$ with $\mu(A) < \delta$, then $\int_A f d\mu < \delta_0$. Note that for any $x \in X$ there exists an integer m such that

$$\frac{f_m(x)}{m} \leq (\underline{f}(x) + \epsilon).$$

Now choose $M > 0$ such that the set

$$Y_0 = \{x \in X : \exists 1 \leq m \leq M \text{ with } f_m(x) \leq m(\underline{f}(x) + \epsilon)\}$$

has measure at least $1 - \delta$. Define G on X by

$$G(x) = \begin{cases} f(x) & x \in Y_0 \\ 0 & x \notin Y_0. \end{cases}$$

Notice that $G \leq f$. Let $b_n = G(T^n x)$, and $a_n = \underline{f}(x) + \epsilon$ (so a_n is independent of n). One can easily check that the sequences $\{a_n\}$ and $\{b_n\}$ satisfy the hypothesis of Lemma 2.1.1 with $M > 0$ as above. Hence for any $M > N$, one has

$$G(x) + \cdots + G(T^{N-M-1}x) \leq N(\underline{f}(x) + \epsilon).$$

Integrating both sides yields

$$(N - M) \int_X G(x) d\mu(x) \leq N \left(\int_X \underline{f}(x) d\mu(x) + \epsilon \right).$$

Since $\mu(X \setminus Y_0) < \delta$, then $\nu(X \setminus Y_0) = \int_{X \setminus Y_0} f(x) d\mu(x) < \delta_0$. Hence,

$$\begin{aligned} \int_X f(x) d\mu(x) &= \int_X G(x) d\mu(x) + \int_{X \setminus Y_0} f(x) d\mu(x) \\ &\leq \frac{N}{N - M} \int_X (\underline{f}(x) + \epsilon) d\mu(x) + \delta_0. \end{aligned}$$

Now, let first $N \rightarrow \infty$, then $\delta \rightarrow 0$ (and hence $\delta_0 \rightarrow 0$), and finally $\epsilon \rightarrow 0$, one gets

$$\int_X f(x) d\mu(x) \leq \int_X \underline{f}(x) d\mu(x).$$

This shows that

$$\int_X \underline{f} d\mu \geq \int_X f d\mu \geq \int_X \bar{f} d\mu,$$

hence, $\overline{f} = \underline{f} = f^*$ a.e., and f^* is T -invariant. In case T is ergodic, then the T -invariance of f^* implies that f^* is a constant a.e. Therefore,

$$f^*(x) = \int_X f^*(y) d\mu(y) = \int_X f(y) d\mu(y).$$

□

Remarks

(1) Let us study further the limit f^* in the case that T is not ergodic. Let \mathcal{I} be the sub- σ -algebra of \mathcal{F} consisting of all T -invariant subsets $A \in \mathcal{F}$. Notice that if $f \in L^1(\mu)$, then the *conditional expectation* of f given \mathcal{I} (denoted by $E_\mu(f|\mathcal{I})$), is the unique a.e. \mathcal{I} -measurable $L^1(\mu)$ function with the property that

$$\int_A f(x) d\mu(x) = \int_A E_\mu(f|\mathcal{I})(x) d\mu(x)$$

for all $A \in \mathcal{I}$ i.e., $T^{-1}A = A$. We claim that $f^* = E_\mu(f|\mathcal{I})$. Since the limit function f^* is T -invariant, it follows that f^* is \mathcal{I} -measurable. Furthermore, for any $A \in \mathcal{I}$, by the ergodic theorem and the T -invariance of 1_A ,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} (f 1_A)(T^i x) = 1_A(x) \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) = 1_A(x) f^*(x) \text{ a.e.}$$

and

$$\int_X f 1_A(x) d\mu(x) = \int_X f^* 1_A(x) d\mu(x).$$

This shows that $f^* = E_\mu(f|\mathcal{I})$.

(2) Suppose T is ergodic and measure preserving with respect to μ , and let ν be a probability measure which is equivalent to μ (i.e. μ and ν have the same sets of measure zero so $\mu(A) = 0$ if and only if $\nu(A) = 0$), then for every $f \in L^1(\mu)$ one has ν a.e.

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i(x)) = \int_X f d\mu.$$

Exercise 2.1.1 (*Kac's Lemma*) Let T be a measure preserving and ergodic transformation on a probability space (X, \mathcal{F}, μ) . Let A be a measurable

subset of X of positive μ measure, and denote by n the first return time map and let T_A be the induced transformation of T on A (see section 1.5). Prove that

$$\int_A n(x) \, d\mu = 1.$$

Conclude that $n(x) \in L^1(A, \mu_A)$, and that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} n(T_A^i(x)) = \frac{1}{\mu(A)},$$

almost everywhere on A .

Exercise 2.1.2 Let $\beta = \frac{1 + \sqrt{5}}{2}$, and consider the transformation $T_\beta : [0, 1) \rightarrow [0, 1)$, given by $T_\beta x = \beta x \bmod(1) = \beta x - [\beta x]$. Define b_1 on $[0, 1)$ by

$$b_1(x) = \begin{cases} 0 & \text{if } 0 \leq x < 1/\beta \\ 1 & \text{if } 1/\beta \leq x < 1, \end{cases}$$

Fix $k \geq 0$. Find the a.e. value (with respect to Lebesgue measure) of the following limit

$$\lim_{n \rightarrow \infty} \frac{1}{n} \#\{1 \leq i \leq n : b_i = 0, b_{i+1} = 0, \dots, b_{i+k} = 0\}.$$

Exercise 2.1.3 Let (X, \mathcal{F}, μ) be a probability space and $f \in L^1(\mu)$. Suppose $\{T_t : t \in \mathbb{R}\}$ is a family of transformations $T_t : X \rightarrow X$ satisfying

- (i) $T_0 = \text{id}_X$ and $T_{t+s} = T_t \circ T_s$
- (ii) T_t is measurable, measure preserving and ergodic w.r.t. μ .
- (iii) The map $G : X \times \mathbb{R} \rightarrow X$ given by $G(x, t) = f(T_t(x))$ is measurable, where $X \times \mathbb{R}$ is endowed with the product σ algebra $\mathcal{F} \times \mathcal{B}$ and product measure $\mu \times \lambda$, with \mathcal{B} the Borel σ -algebra on \mathbb{R} and λ is Lebesgue measure.

(a) Show that for all $s \geq 0$,

$$\int_0^s \int_X f(T_t(x)) d\mu(x) d\lambda(t) = \int_X \int_0^s f(T_t(x)) d\lambda(t) d\mu(x) = s \int_X f(x) d\mu(x).$$

(b) Show that for all $s \geq 0$, $\int_0^s f(T_t(x)) d\lambda(t) < \infty$ μ a.e.

(c) Define $F : X \rightarrow \mathbb{R}$ by $F(x) = \int_0^1 f(T_t(x)) d\lambda(t)$, and consider the transformation T_1 corresponding to $t = 1$. Show that for any $n \geq 1$ one has

$$\sum_{k=0}^{n-1} F(T_1^k(x)) = \int_0^n f(T_t(x)) d\lambda(t),$$

$$\text{and } \int_X F d\mu = \int_X f d\mu.$$

(d) Show that for μ a.e. x one has

$$\lim_{n \rightarrow \infty} \frac{1}{n} \int_0^n f(T_t(x)) d\lambda(t) = \int_X f d\mu.$$

Using the Ergodic Theorem, one can give yet another characterization of ergodicity.

Corollary 2.1.1 *Let (X, \mathcal{F}, μ) be a probability space, and $T : X \rightarrow X$ a measure preserving transformation. Then, T is ergodic if and only if for all $A, B \in \mathcal{F}$, one has*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}A \cap B) = \mu(A)\mu(B). \quad (2.1)$$

Proof. Suppose T is ergodic, and let $A, B \in \mathcal{F}$. Since the indicator function $1_A \in L^1(X, \mathcal{F}, \mu)$, by the ergodic theorem one has

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_A(T^i x) = \int_X 1_A(x) d\mu(x) = \mu(A) \text{ a.e.}$$

Then,

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_{T^{-i}A \cap B}(x) &= \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_{T^{-i}A}(x) 1_B(x) \\ &= 1_B(x) \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_A(T^i x) \\ &= 1_B(x) \mu(A) \quad \text{a.e.} \end{aligned}$$

Since for each n , the function $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_{T^{-i}A \cap B}$ is dominated by the constant function 1, it follows by the dominated convergence theorem that

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}A \cap B) &= \int_X \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_{T^{-i}A \cap B}(x) d\mu(x) \\ &= \int_X 1_B \mu(A) d\mu(x) \\ &= \mu(A) \mu(B). \end{aligned}$$

Conversely, suppose (2.1) holds for every $A, B \in \mathcal{F}$. Let $E \in \mathcal{F}$ be such that $T^{-1}E = E$ and $\mu(E) > 0$. By invariance of E , we have $\mu(T^{-i}E \cap E) = \mu(E)$, hence

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}E \cap E) = \mu(E).$$

On the other hand, by (2.1)

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}E \cap E) = \mu(E)^2.$$

Hence, $\mu(E) = \mu(E)^2$. Since $\mu(E) > 0$, this implies $\mu(E) = 1$. Therefore, T is ergodic. \square

To show ergodicity one needs to verify equation (2.1) for sets A and B belonging to a generating semi-algebra only as the next proposition shows.

Proposition 2.1.1 *Let (X, \mathcal{F}, μ) be a probability space, and \mathcal{S} a generating semi-algebra of \mathcal{F} . Let $T : X \rightarrow X$ be a measure preserving transformation. Then, T is ergodic if and only if for all $A, B \in \mathcal{S}$, one has*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}A \cap B) = \mu(A) \mu(B). \quad (2.2)$$

Proof. We only need to show that if (2.2) holds for all $A, B \in \mathcal{S}$, then it holds for all $A, B \in \mathcal{F}$. Note that (2.2) holds for elements in the algebra generated by \mathcal{S} . Let $\epsilon > 0$, and $A, B \in \mathcal{F}$. Then, by Lemma 1.2.1 (in Subsection 1.2) there exist sets A_0, B_0 each of which is a finite disjoint union of elements of \mathcal{S} such that

$$\mu(A\Delta A_0) < \epsilon, \quad \text{and} \quad \mu(B\Delta B_0) < \epsilon.$$

Since,

$$(T^{-i}A \cap B)\Delta(T^{-i}A_0 \cap B_0) \subseteq (T^{-i}A\Delta T^{-i}A_0) \cup (B\Delta B_0),$$

it follows that

$$\begin{aligned} |\mu(T^{-i}A \cap B) - \mu(T^{-i}A_0 \cap B_0)| &\leq \mu[(T^{-i}A \cap B)\Delta(T^{-i}A_0 \cap B_0)] \\ &\leq \mu(T^{-i}A\Delta T^{-i}A_0) + \mu(B\Delta B_0) \\ &< 2\epsilon. \end{aligned}$$

Further,

$$\begin{aligned} |\mu(A)\mu(B) - \mu(A_0)\mu(B_0)| &\leq \mu(A)|\mu(B) - \mu(B_0)| + \mu(B_0)|\mu(A) - \mu(A_0)| \\ &\leq |\mu(B) - \mu(B_0)| + |\mu(A) - \mu(A_0)| \\ &\leq \mu(B\Delta B_0) + \mu(A\Delta A_0) \\ &< 2\epsilon. \end{aligned}$$

Hence,

$$\begin{aligned} &\left| \left(\frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}A \cap B) - \mu(A)\mu(B) \right) - \left(\frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}A_0 \cap B_0) - \mu(A_0)\mu(B_0) \right) \right| \\ &\leq \frac{1}{n} \sum_{i=0}^{n-1} |\mu(T^{-i}A \cap B) + \mu(T^{-i}A_0 \cap B_0)| - |\mu(A)\mu(B) - \mu(A_0)\mu(B_0)| \\ &< 4\epsilon. \end{aligned}$$

Therefore,

$$\lim_{n \rightarrow \infty} \left[\frac{1}{n} \sum_{i=0}^{n-1} \mu(T^{-i}A \cap B) - \mu(A)\mu(B) \right] = 0.$$

□

Theorem 2.1.2 Suppose μ_1 and μ_2 are probability measures on (X, \mathcal{F}) , and $T : X \rightarrow X$ is measurable and measure preserving with respect to μ_1 and μ_2 . Then,

- (i) if T is ergodic with respect to μ_1 , and μ_2 is absolutely continuous with respect to μ_1 , then $\mu_1 = \mu_2$,
- (ii) if T is ergodic with respect to μ_1 and μ_2 , then either $\mu_1 = \mu_2$ or μ_1 and μ_2 are singular with respect to each other.

Proof. (i) Suppose T is ergodic with respect to μ_1 and μ_2 is absolutely continuous with respect to μ_1 . For any $A \in \mathcal{F}$, by the ergodic theorem for a.e. x one has

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_A(T^i x) = \mu_1(A).$$

Let

$$C_A = \left\{ x \in X : \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_A(T^i x) = \mu_1(A) \right\},$$

then $\mu_1(C_A) = 1$, and by absolute continuity of μ_2 one has $\mu_2(C_A) = 1$. Since T is measure preserving with respect to μ_2 , for each $n \geq 1$ one has

$$\frac{1}{n} \sum_{i=0}^{n-1} \int_X 1_A(T^i x) d\mu_2(x) = \mu_2(A).$$

On the other hand, by the dominated convergence theorem one has

$$\lim_{n \rightarrow \infty} \int_X \frac{1}{n} \sum_{i=0}^{n-1} 1_A(T^i x) d\mu_2(x) = \int_X \mu_1(A) d\mu_2(x).$$

This implies that $\mu_1(A) = \mu_2(A)$. Since $A \in \mathcal{F}$ is arbitrary, we have $\mu_1 = \mu_2$.

(ii) Suppose T is ergodic with respect to μ_1 and μ_2 . Assume that $\mu_1 \neq \mu_2$. Then, there exists a set $A \in \mathcal{F}$ such that $\mu_1(A) \neq \mu_2(A)$. For $i = 1, 2$ let

$$C_i = \left\{ x \in X : \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^{n-1} 1_A(T^j x) = \mu_i(A) \right\}.$$

By the ergodic theorem $\mu_i(C_i) = 1$ for $i = 1, 2$. Since $\mu_1(A) \neq \mu_2(A)$, then $C_1 \cap C_2 = \emptyset$. Thus μ_1 and μ_2 are supported on disjoint sets, and hence μ_1 and μ_2 are mutually singular. \square

We end this subsection with a short discussion that the assumption of ergodicity is not very restrictive. Let T be a transformation on the probability space (X, \mathcal{F}, μ) , and suppose T is measure preserving but not necessarily ergodic. We assume that X is a complete separable metric space, and \mathcal{F} the corresponding Borel σ -algebra (in order to make sure that the conditional expectation is well-defined a.e.). Let \mathcal{I} be the sub- σ -algebra of T -invariant measurable sets. We can decompose μ into T -invariant ergodic components in the following way. For $x \in X$, define a measure μ_x on \mathcal{F} by

$$\mu_x(A) = E_\mu(1_A | \mathcal{I})(x).$$

Then, for any $f \in L^1(X, \mathcal{F}, \mu)$,

$$\int_X f(y) d\mu_x(y) = E_\mu(f | \mathcal{I})(x).$$

Note that

$$\mu(A) = \int_X E_\mu(1_A | \mathcal{I})(x) d\mu(x) = \int_X \mu_x(A) d\mu(x),$$

and that $E_\mu(1_A | \mathcal{I})(x)$ is T -invariant. We show that μ_x is T -invariant and ergodic for a.e. $x \in X$. So let $A \in \mathcal{F}$, then for a.e. x

$$\mu_x(T^{-1}A) = E_\mu(1_A \circ T | \mathcal{I})(x) = E_\mu(1_A | \mathcal{I})(Tx) = E_\mu(1_A | \mathcal{I})(x) = \mu_x(A).$$

Now, let $A \in \mathcal{F}$ be such that $T^{-1}A = A$. Then, 1_A is T -invariant, and hence \mathcal{I} -measurable. Then,

$$\mu_x(A) = E_\mu(1_A | \mathcal{I})(x) = 1_A(x) \text{ a.e.}$$

Hence, for a.e. x and for any $B \in \mathcal{F}$,

$$\mu_x(A \cap B) = E_\mu(1_A 1_B | \mathcal{I})(x) = 1_A(x) E_\mu(1_B | \mathcal{I})(x) = \mu_x(A) \mu_x(B).$$

In particular, if $A = B$, then the latter equality yields $\mu_x(A) = \mu_x(A)^2$ which implies that for a.e. x , $\mu_x(A) = 0$ or 1 . Therefore, μ_x is ergodic. (One in fact needs to work a little harder to show that one can find a set N of μ -measure zero, such that for any $x \in X \setminus N$, and any T -invariant set A , one has $\mu_x(A) = 0$ or 1 . In the above analysis the a.e. set depended on the choice of A . Hence, the above analysis is just a rough sketch of the proof of what is called *the ergodic decomposition* of measure preserving transformations.)

2.2 The Mean Ergodic Theorem

In the previous section, we studied the pointwise behaviour of the ergodic averages $\frac{1}{n} \sum_{i=0}^{n-1} f(T^i x)$ for $f \in L^1(\mu)$. In this section, we will restrict our attention to $f \in L^2(\mu)$, and study the $L^2(\mu)$ convergence of the ergodic averages. The result that we will be proving is due to von Neumann (1932). Before we start let us recall few facts over the Hilbert space $L^2(\mu)$ and the operator U_T defined in section 1.7.

We begin with the space $L^2(\mu) = L^2(X, \mathcal{F}, \mu)$ which is a Hilbert space equipped with the inner product $(f, g) = \int f \bar{g} d\mu$, where \bar{g} is the complex conjugate of g . The inner product induces the $L^2(\mu)$ norm defined by $\|f\|_2 = (f, f)^{1/2}$. If S is a closed linear subspace of $L^2(\mu)$, then the **orthogonal complement** of S is defined by

$$S^\perp = \{h \in L^2(\mu) : (h, g) = 0 \text{ for all } g \in S\}.$$

The following well known Theorem states that each element $f \in L^2(\mu)$ can be uniquely written as a sum $f = g + h$, where $g \in S$ and $h \in S^\perp$. We state it without a proof, and refer the reader to any standard book on functional analysis.

Theorem 2.2.1 Let S be a closed linear subspace of $L^2(\mu)$. Then for any element $f \in L^2(\mu)$, there exists a unique element $g \in S$ satisfying

$$\inf\{\|f - f'\|_2 : f' \in S\} = \|f - g\|_2.$$

Furthermore, if $P : L^2(\mu) \rightarrow S$ is defined by $P(f) = g$, then every element $f \in L^2(\mu)$ can be written uniquely as $f = P(f) + h$, where $h \in S^\perp$. The transformation P is called the **orthogonal projection** of $L^2(\mu)$ onto S .

Now suppose (X, \mathcal{F}, μ) is a probability space and $T : X \rightarrow X$ is a measure preserving transformation. Consider the operator U_T defined in section 1.7, but restricted to $L^2(\mu)$,

$$U_T(f) = f \circ T, \text{ for } f \in L^2(\mu).$$

Note that U_T is an isometry since $(U_T(f), U_T(f)) = (f, f)$. Associated with U_T , one defines the *adjoint operator* $U_T^* : L^2(\mu) \rightarrow L^2(\mu)$ satisfying

$$(U_T(f), g) = (f, U_T^*(g)), \text{ for all } f, g \in L^2(\mu).$$

Since U_T is an isometry, one gets that $U_T^*U_T = I_{L^2(\mu)}$, where $I_{L^2(\mu)}$ is the identity operator. Consider the set

$$\mathcal{I} = \{f \in L^2(\mu) : f = U_T(f) = f \circ T\}.$$

It is easy to check that \mathcal{I} is a closed linear subspace of $L^2(\mu)$. By Theorem 2.2.1, we can write each element $f \in L^2(\mu)$ as $f = g + h$, where $g = Pf \in \mathcal{I}$ and $h \in \mathcal{I}^\perp$. In the following Lemma, we identify explicitly the elements of \mathcal{I}^\perp .

Lemma 2.2.1 Let $B = \{U_T(g) - g : g \in L^2(\mu)\}$. Then $\overline{B} = \mathcal{I}^\perp$, where \overline{B} is the closure of B in the $L^2(\mu)$ norm.

Proof. Equivalently, we will show that $\overline{B}^\perp = \mathcal{I}$. Let $f \in \mathcal{I}$, then $U_T(f) = f$ and for any $(U_T(g) - g) \in B$, we have

$$(f, U_T(g) - g) = (U_T(f), U_T(g)) - (f, g) = (f, g) - (f, g) = 0.$$

Thus f is orthogonal to every element of B . We show this is true for elements of \overline{B} as well. Let $h \in \overline{B}$, and $(h_i = U_T(g_i) - g_i)$ a sequence in B converging in $L^2(\mu)$ to h . Then, $(f, h_i) = 0$ for all i , and by Cauchy Schwartz inequality, $(f, h) = \lim_{i \rightarrow \infty} (f, h_i) = 0$. Thus $f \in \overline{B}^\perp$. Conversely, suppose $f \in \overline{B}^\perp$. For every $g \in L^2(\mu)$ $((f, U_T(g) - g) = 0)$, we have

$$(f, g) = (f, U_T(g)) = (U_T^*(f), g).$$

This implies that $f = U_T^*(f)$. We now show that $f = U_T(f)$. To this end consider

$$\begin{aligned} \|U_T(f) - f\|_2^2 &= (U_T(f) - f, U_T(f) - f) \\ &= \|U_T(f)\|_2^2 - (U_T(f), f) - (f, U_T(f)) + \|f\|_2^2 \\ &= \|f\|_2^2 - (f, U_T^*(f)) - (U_T^*(f), f) + \|f\|_2^2 = 0 \end{aligned}$$

Thus, $f \in \mathcal{I}$. \square

We are now ready to prove the Mean Ergodic Theorem.

Theorem 2.2.2 Let (X, \mathcal{F}, μ, T) be a measure preserving system, and let P_T denote the orthogonal projection onto the closed subspace \mathcal{I} . Then, for

any $f \in L^2(\mu)$ the sequence $(\frac{1}{n} \sum_{i=0}^{n-1} f \circ T^i)$ converges in $L^2(\mu)$ to $P_T(f)$.

Proof. By Theorem 2.2.1 and Lemma 2.2.1, any $f \in L^2(\mu)$ can be written as $f = P_T(f) + h$ with $h \in \overline{B}$. Now $P_T(f) \in \mathcal{I}$, hence $P_T(f) \circ T = U_T(P_T(f)) = P_T(f)$. This implies that $\frac{1}{n} \sum_{i=0}^{n-1} P_T(f) \circ T^i$ converges in $L^2(\mu)$ to $P_T(f)$. Now let $(h_j = U_T(g_j) - g_j)$ be a sequence in B converging to h in $L^2(\mu)$. Note that

$$\frac{1}{n} \sum_{i=0}^{n-1} h_j \circ T^i = \frac{1}{n} (g_j \circ T^n - g_j) = \frac{1}{n} (U_T^n(g_j) - g_j).$$

Thus,

$$\begin{aligned} \left\| \frac{1}{n} \sum_{i=0}^{n-1} h_j \circ T^i \right\|_2 &= \left\| \frac{1}{n} \sum_{i=0}^{n-1} U_T^i(h_j) \right\|_2 \\ &= \frac{1}{n} \|U_T^n(g_j) - g_j\|_2 \\ &\leq \frac{2}{n} \|g_j\|_2. \end{aligned}$$

This shows that $\frac{1}{n} \sum_{i=0}^{n-1} h_j \circ T^i$ converges in $L^2(\mu)$ to 0 for all j . Finally, for any j sufficiently large,

$$\begin{aligned} \left\| \frac{1}{n} \sum_{i=0}^{n-1} h \circ T^i \right\|_2 &\leq \frac{1}{n} \sum_{i=0}^{n-1} \|(h - h_j) \circ T^i\|_2 + \left\| \frac{1}{n} \sum_{i=0}^{n-1} h_j \circ T^i \right\|_2 \\ &= \frac{1}{n} \sum_{i=0}^{n-1} \|h - h_j\|_2 + \left\| \frac{1}{n} \sum_{i=0}^{n-1} h_j \circ T^i \right\|_2 \\ &= \|h - h_j\|_2 + \left\| \frac{1}{n} \sum_{i=0}^{n-1} h_j \circ T^i \right\|_2. \end{aligned}$$

Taking the limit as $n \rightarrow \infty$ we see that $\frac{1}{n} \sum_{i=0}^{n-1} h \circ T^i$ converges to 0 in $L^2(\mu)$.

Since

$$\frac{1}{n} \sum_{i=0}^{n-1} f \circ T^i = \frac{1}{n} \sum_{i=0}^{n-1} P_T(f) \circ T^i + \frac{1}{n} \sum_{i=0}^{n-1} h \circ T^i,$$

we get the required result. \square

2.3 Mixing

As a corollary to the Pointwise Ergodic Theorem we found a new definition of ergodicity; namely, asymptotic average independence. Based on the same idea, we now define other notions of *weak* independence that are stronger than ergodicity.

Definition 2.3.1 *Let (X, \mathcal{F}, μ) be a probability space, and $T : X \rightarrow X$ a measure preserving transformation. Then,*

(i) *T is weakly mixing if for all $A, B \in \mathcal{F}$, one has*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} |\mu(T^{-i}A \cap B) - \mu(A)\mu(B)| = 0. \quad (2.3)$$

(ii) *T is strongly mixing if for all $A, B \in \mathcal{F}$, one has*

$$\lim_{n \rightarrow \infty} \mu(T^{-i}A \cap B) = \mu(A)\mu(B). \quad (2.4)$$

Notice that strongly mixing implies weakly mixing, and weakly mixing implies ergodicity. This follows from the simple fact that if $\{a_n\}$ is a sequence of real numbers such that $\lim_{n \rightarrow \infty} a_n = 0$, then $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} |a_i| = 0$, and

hence $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} a_i = 0$. Furthermore, if $\{a_n\}$ is a bounded sequence of non-negative real numbers, then the following are equivalent (see [W] for the proof):

(i) $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} |a_i| = 0$

(ii) $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} |a_i|^2 = 0$

(iii) there exists a subset J of the integers of density zero, i.e.

$$\lim_{n \rightarrow \infty} \frac{1}{n} \#(\{0, 1, \dots, n-1\} \cap J) = 0,$$

such that $\lim_{n \rightarrow \infty, n \notin J} a_n = 0$.

Using this one can give three equivalent characterizations of weakly mixing transformations, can you state them?

Exercise 2.3.1 Let (X, \mathcal{F}, μ) be a probability space, and $T : X \rightarrow X$ a measure preserving transformation. Let \mathcal{S} be a generating semi-algebra of \mathcal{F} .

- (a) Show that if equation (2.3) holds for all $A, B \in \mathcal{S}$, then T is weakly mixing.
- (b) Show that if equation (2.4) holds for all $A, B \in \mathcal{S}$, then T is strongly mixing.

Exercise 2.3.2 Consider the one or two-sided Bernoulli shift T as given in Example 1.3.6 and Example 1.8.2. Show that T is strongly mixing.

Exercise 2.3.3 Let (X, \mathcal{F}, μ) be a probability space, and let $T : X \rightarrow X$ measure preserving and strongly mixing. Consider the probability space (Y, \mathcal{G}, ν) , where

$$Y = X \times \{0\} \cup X \times \{1\} \cup X \times \{2\},$$

\mathcal{G} the σ -algebra generated by sets of the form $A \times \{i\}$ with $A \in \mathcal{F}$, $i = 0, 1, 2$, and ν the measure given by $\nu(A \times \{i\}) = \frac{1}{3}\mu(A)$. Define $S : Y \rightarrow Y$ by $S(x, 0) = (x, 1)$, $S(x, 1) = (x, 2)$ and $S(x, 2) = (Tx, 0)$.

- (a) Show that S is measure preserving and ergodic with respect to ν .
- (b) Show that $S^2 = S \circ S$ is ergodic, but $S^3 = S \circ S \circ S$ is not ergodic.
- (c) Show that S is not strongly mixing.

Exercise 2.3.4 Let (X, \mathcal{F}, μ) be a probability space, and $T : X \rightarrow X$ a measure preserving transformation. Consider the transformation $T \times T$ defined on $(X \times X, \mathcal{F} \times \mathcal{F}, \mu \times \mu)$ by $T \times T(x, y) = (Tx, Ty)$.

- (a) Show that $T \times T$ is measure preserving with respect to $\mu \times \mu$.
- (b) Show that $T \times T$ is ergodic, if and only if T is weakly mixing.

Exercise 2.3.5 Let (X, \mathcal{F}, μ, T) be a measure preserving dynamical system.

- (a) Show that the following are equivalent.
- (i) T is weakly mixing.
 - (ii) For any ergodic dynamical system (Y, \mathcal{G}, ν, S) , the measure preserving dynamical system $(X \times Y, \mathcal{F} \times \mathcal{G}, \mu \times \nu, T \times S)$ is ergodic, where $\mathcal{F} \times \mathcal{G}$ is the product σ -algebra, $\mu \times \nu$ is the product measure, and $T \times S(x, y) = (Tx, Sy)$.
- (b) Show that if $T = T_\theta = x + \theta \pmod{1}$ is an irrational rotation on $[0, 1)$, then T_θ is **not** weakly mixing with respect to λ , where λ is the normalized Lebesgue measure on $[0, 1)$.

Chapter 3

Measure Preserving Isomorphisms and Factor Maps

3.1 Measure Preserving Isomorphisms

Given a measure preserving transformation T on a probability space (X, \mathcal{F}, μ) , we call the quadruple (X, \mathcal{F}, μ, T) a *dynamical system*. Now, given two dynamical systems (X, \mathcal{F}, μ, T) and (Y, \mathcal{C}, ν, S) , what should we mean by: *these systems are the same*? On each space there are two important structures:

- (1) The measure structure given by the σ -algebra and the probability measure. Note, that in this context, sets of measure zero can be ignored.
- (2) The dynamical structure, given by a measure preserving transformation.

So our notion of *being the same* must mean that we have a map

$$\psi : (X, \mathcal{F}, \mu, T) \rightarrow (Y, \mathcal{C}, \nu, S)$$

satisfying

- (i) ψ is one-to-one and onto a.e. By this we mean, that if we remove a (suitable) set N_X of measure 0 in X , and a (suitable) set N_Y of measure 0 in Y , the map $\psi : X \setminus N_X \rightarrow Y \setminus N_Y$ is a bijection.
- (ii) ψ is measurable, i.e., $\psi^{-1}(C) \in \mathcal{F}$, for all $C \in \mathcal{C}$.

- (iii) ψ preserves the measures: $\nu = \mu \circ \psi^{-1}$, i.e., $\nu(C) = \mu(\psi^{-1}(C))$ for all $C \in \mathcal{C}$.

Finally, we should have that

- (iv) ψ preserves the dynamics of T and S , i.e., $\psi \circ T = S \circ \psi$, which is the same as saying that the following diagram commutes.

$$\begin{array}{ccc}
 N & \xrightarrow{T} & N \\
 \psi \downarrow & & \downarrow \psi \\
 N' & \xrightarrow{S} & N'
 \end{array}$$

This means that T -orbits are mapped to S -orbits:

$$\begin{array}{ccccccc}
 x & \rightarrow & Tx & \rightarrow & T^2x & \rightarrow & \cdots \rightarrow & T^n x & \rightarrow \\
 \downarrow & & \downarrow & & \downarrow & & \downarrow & & \downarrow \\
 \psi(x) & \rightarrow & S(\psi(x)) & \rightarrow & S^2(\psi(x)) & \rightarrow & \cdots \rightarrow & S^n(\psi(x)) & \rightarrow
 \end{array}$$

Definition 3.1.1 Two dynamical systems (X, \mathcal{F}, μ, T) and (Y, \mathcal{C}, ν, S) are isomorphic if there exist measurable sets $N \subset X$ and $M \subset Y$ with $\mu(X \setminus N) = \nu(Y \setminus M) = 0$ and $T(N) \subset N$, $S(M) \subset M$, and finally if there exists a measurable map $\psi : N \rightarrow M$ such that (i)–(iv) are satisfied.

Exercise 3.1.1 Suppose (X, \mathcal{F}, μ, T) and (Y, \mathcal{C}, ν, S) are two isomorphic dynamical systems. Show that

- T is ergodic if and only if S is ergodic.
- T is weakly mixing if and only if S is weakly mixing.
- T is strongly mixing if and only if S is strongly mixing.

Example 3.1.1 Let $K = \{z \in \mathbb{C} : |z| = 1\}$ be equipped with the Borel σ -algebra \mathcal{B} on K , and Haar measure (i.e., normalized Lebesgue measure on the unit circle). Define $S : K \rightarrow K$ by $Sz = z^2$; equivalently $Se^{2\pi i\theta} = e^{2\pi i(2\theta)}$. One can easily check that S is measure preserving. In fact, the map S is isomorphic to the map T on $([0, 1), \mathcal{B}, \lambda)$ given by $Tx = 2x \pmod{1}$ (see Example 1.3.2 in Subsection 1.3, and Example 1.8.4 in Subsection 1.8). Define a map $\phi : [0, 1) \rightarrow K$ by $\phi(x) = e^{2\pi ix}$. We leave it to the reader to check that ϕ is a measurable isomorphism, i.e., ϕ is a measurable and measure preserving bijection such that $S\phi(x) = \phi(Tx)$ for all $x \in [0, 1)$.

Example 3.1.2 Consider $([0, 1), \mathcal{B}, \lambda)$, the unit interval with the Lebesgue σ -algebra, and Lebesgue measure. Let $T : [0, 1) \rightarrow [0, 1)$ be given by $Tx = Nx - \lfloor Nx \rfloor$. Iterations of T generate the N -adic expansion of points in the unit interval. Let $Y := \{0, 1, \dots, N-1\}^{\mathbb{N}}$, the set of all sequences $(y_n)_{n \geq 1}$, with $y_n \in \{0, 1, \dots, N-1\}$ for $n \geq 1$. We now construct an isomorphism between $([0, 1), \mathcal{B}, \lambda, T)$ and (Y, \mathcal{F}, μ, S) , where \mathcal{F} is the σ -algebra generated by the cylinders, and μ the uniform product measure defined on cylinders by

$$\mu(\{(y_i)_{i \geq 1} \in Y : y_1 = a_1, y_2 = a_2, \dots, y_n = a_n\}) = \frac{1}{N^n},$$

for any $(a_1, a_2, a_3, \dots) \in Y$, and where S is the left shift.

Define $\psi : [0, 1) \rightarrow Y = \{0, 1, \dots, N-1\}^{\mathbb{N}}$ by

$$\psi : x = \sum_{k=1}^{\infty} \frac{a_k}{N^k} \mapsto (a_k)_{k \geq 1},$$

where $\sum_{k=1}^{\infty} a_k/N^k$ is the N -adic expansion of x (for example if $N = 2$ we get the binary expansion, and if $N = 10$ we get the decimal expansion). Let

$$C(i_1, \dots, i_n) = \{(y_i)_{i \geq 1} \in Y : y_1 = i_1, \dots, y_n = i_n\}.$$

In order to see that ψ is an isomorphism one needs to verify measurability and measure preservingness on cylinders:

$$\psi^{-1}(C(i_1, \dots, i_n)) = \left[\frac{i_1}{N} + \frac{i_2}{N^2} + \dots + \frac{i_n}{N^n}, \frac{i_1}{N} + \frac{i_2}{N^2} + \dots + \frac{i_n + 1}{N^n} \right)$$

and

$$\lambda(\psi^{-1}(C(i_1, \dots, i_n))) = \frac{1}{N^n} = \mu(C(i_1, \dots, i_n)).$$

Note that

$$\mathcal{N} = \{(y_i)_{i \geq 1} \in Y : \text{there exists a } k \geq 1 \text{ such that } y_i = N-1 \text{ for all } i \geq k\}$$

is a subset of Y of measure 0. Setting $\tilde{Y} = Y \setminus \mathcal{N}$, then $\psi : [0, 1) \rightarrow \tilde{Y}$ is a bijection, since every $x \in [0, 1)$ has a unique N -adic expansion **generated** by T . Finally, it is easy to see that $\psi \circ T = S \circ \psi$.

Exercise 3.1.2 Consider $([0, 1)^2, \mathcal{B} \times \mathcal{B}, \lambda \times \lambda)$, where $\mathcal{B} \times \mathcal{B}$ is the product Lebesgue σ -algebra, and $\lambda \times \lambda$ is the product Lebesgue measure. Let $T : [0, 1)^2 \rightarrow [0, 1)^2$ be given by

$$T(x, y) = \begin{cases} (2x, \frac{1}{2}y), & 0 \leq x < \frac{1}{2} \\ (2x-1, \frac{1}{2}(y+1)), & \frac{1}{2} \leq x < 1. \end{cases}$$

Show that T is isomorphic to the two-sided Bernoulli shift S on $(\{0, 1\}^{\mathbb{Z}}, \mathcal{F}, \mu)$, where \mathcal{F} is the σ -algebra generated by cylinders of the form

$$\Delta = \{x_{-k} = a_{-k}, \dots, x_{\ell} = a_{\ell} : a_i \in \{0, 1\}, i = -k, \dots, \ell\}, \quad k, \ell \geq 0,$$

and μ the product measure with weights $(\frac{1}{2}, \frac{1}{2})$ (so $\mu(\Delta) = (\frac{1}{2})^{k+\ell+1}$).

Exercise 3.1.3 Let $G = \frac{1 + \sqrt{5}}{2}$, so that $G^2 = G + 1$. Consider the set

$$X = [0, \frac{1}{G}) \times [0, 1) \cup [\frac{1}{G}, 1) \times [0, \frac{1}{G}),$$

endowed with the product Borel σ -algebra. Define the transformation

$$\mathcal{T}(x, y) = \begin{cases} (Gx, \frac{y}{G}), & (x, y) \in [0, \frac{1}{G}) \times [0, 1) \\ (Gx-1, \frac{1+y}{G}), & (x, y) \in [\frac{1}{G}, 1) \times [0, \frac{1}{G}). \end{cases}$$

- (a) Show that \mathcal{T} is measure preserving with respect to normalized Lebesgue measure on X .
- (b) Now let $\mathcal{S} : [0, 1) \times [0, 1) \rightarrow [0, 1) \times [0, 1)$ be given by

$$\mathcal{S}(x, y) = \begin{cases} (Gx, \frac{y}{G}), & (x, y) \in [0, \frac{1}{G}) \times [0, 1) \\ (G^2x - G, \frac{G+y}{G^2}), & (x, y) \in [\frac{1}{G}, 1) \times [0, 1). \end{cases}$$

Show that \mathcal{S} is measure preserving with respect to normalized Lebesgue measure on $[0, 1) \times [0, 1)$.

- (c) Let $Y = [0, 1) \times [0, \frac{1}{G})$, and let U be the induced transformation of \mathcal{T} on Y , i.e., for $(x, y) \in Y$, $U(x, y) = \mathcal{T}^{n(x, y)}$, where $n(x, y) = \inf\{n \geq 1 : \mathcal{T}^n(x, y) \in Y\}$. Show that the map $\phi : [0, 1) \times [0, 1) \rightarrow Y$ given by

$$\phi(x, y) = (x, \frac{y}{G})$$

defines an isomorphism from \mathcal{S} to U , where Y has the induced measure structure; see Section 1.5.

Exercise 3.1.4

Consider the measurable space $(\mathbb{R}, \mathcal{B})$, where \mathbb{R} is the real line and \mathcal{B} is the Borel σ -algebra. Define a transformation $T : \mathbb{R} \rightarrow \mathbb{R}$ by

$$Tx = \frac{1}{2}\left(x - \frac{1}{x}\right).$$

Consider the measure μ on \mathcal{B} defined by

$$\mu(A) = \int_A \frac{1}{\pi(1+x^2)} d\lambda(x),$$

where λ is Lebesgue measure.

- (a) Show that T is measure preserving with respect to the probability measure μ .

(b) Let $\phi : \mathbb{R} \rightarrow [0, 1)$ be defined by

$$\phi(x) = \frac{1}{\pi} \arctan x + \frac{1}{2}.$$

Show that ϕ is an isomorphism between the dynamical systems $(\mathbb{R}, \mathcal{B}, \mu, T)$ and $([0, 1), \mathcal{B}_{[0,1)}, \lambda_{[0,1)}, S)$, where $\mathcal{B}_{[0,1)} = \mathcal{B} \cap [0, 1)$ is the restriction of the Borel σ -algebra on $[0, 1)$, $\lambda_{[0,1)}$ is the restriction of Lebesgue measure on $[0, 1)$, and $Sx = 2x \bmod 1$.

(c) Show that T is strongly mixing with respect to μ .

3.2 Factor Maps

In the above section, we discussed the notion of isomorphism which describes when two dynamical systems are considered the same. Now, we give a precise definition of what it means for a dynamical system to be a subsystem of another one.

Definition 3.2.1 *Let (X, \mathcal{F}, μ, T) and (Y, \mathcal{C}, ν, S) be two dynamical systems. We say that S is a factor of T if there exist measurable sets $M_1 \in \mathcal{F}$ and $M_2 \in \mathcal{C}$, such that $\mu(M_1) = \nu(M_2) = 1$ and $T(M_1) \subset M_1$, $S(M_2) \subset M_2$, and finally if there exists a measurable and measure preserving map $\psi : M_1 \rightarrow M_2$ which is surjective, and satisfies $\psi(T(x)) = S(\psi(x))$ for all $x \in M_1$. We call ψ a factor map.*

Remark Notice that if ψ is a factor map, then $\mathcal{G} = \psi^{-1}\mathcal{C}$ is a T -invariant sub- σ -algebra of \mathcal{F} , since

$$T^{-1}\mathcal{G} = T^{-1}\psi^{-1}\mathcal{C} = \psi^{-1}S^{-1}\mathcal{C} \subseteq \psi^{-1}\mathcal{C} = \mathcal{G}.$$

Examples. Let T be the Baker's transformation on $([0, 1)^2, \mathcal{B} \times \mathcal{B}, \lambda \times \lambda)$, given by

$$T(x, y) = \begin{cases} (2x, \frac{1}{2}y) & \text{if } 0 \leq x < \frac{1}{2} \\ (2x - 1, \frac{1}{2}(y + 1)) & \text{if } \frac{1}{2} \leq x < 1, \end{cases}$$

and let S be the left shift on $X = \{0, 1\}^{\mathbb{N}}$ with the σ -algebra \mathcal{F} generated by the cylinders, and the uniform product measure μ . Define $\psi : [0, 1) \times [0, 1) \rightarrow X$ by

$$\psi(x, y) = (a_1, a_2, \dots),$$

where $x = \sum_{n=1}^{\infty} a_n/2^n$ is the binary expansion of x . It is easy to check that ψ is a factor map.

Exercise 3.2.1 Let T be the left shift on $X = \{0, 1, 2\}^{\mathbb{N}}$ which is endowed with the σ -algebra \mathcal{F} , generated by the cylinder sets, and the uniform product measure μ giving each symbol probability $1/3$, i.e.,

$$\mu(\{x \in X : x_1 = i_1, x_2 = i_2, \dots, x_n = i_n\}) = \frac{1}{3^n},$$

where $i_1, i_2, \dots, i_n \in \{0, 1, 2\}$.

Let S be the left shift on $Y = \{0, 1\}^{\mathbb{N}}$ which is endowed with the σ -algebra \mathcal{G} , generated by the cylinder sets, and the product measure ν giving the symbol 0 probability $1/3$ and the symbol 1 probability $2/3$, i.e.,

$$\mu(\{y \in Y : y_1 = j_1, y_2 = j_2, \dots, y_n = j_n\}) = \left(\frac{2}{3}\right)^{j_1 + \dots + j_n} \left(\frac{1}{3}\right)^{n - (j_1 + \dots + j_n)},$$

where $j_1, j_2, \dots, j_n \in \{0, 1\}$. Show that S is a factor of T .

Exercise 3.2.2 Show that a factor of an ergodic (weakly mixing/strongly mixing) transformation is also ergodic (weakly mixing/strongly mixing).

3.3 Natural Extensions

Suppose (Y, \mathcal{G}, ν, S) is a *non-invertible* measure-preserving dynamical system. An invertible measure-preserving dynamical system (X, \mathcal{F}, μ, T) is called a *natural extension* of (Y, \mathcal{G}, ν, S) if S is a factor of T and the factor map ψ satisfies $\bigvee_{m=0}^{\infty} T^m \psi^{-1} \mathcal{G} = \mathcal{F}$, where

$$\bigvee_{k=0}^{\infty} T^k \psi^{-1} \mathcal{G}$$

is the smallest σ -algebra containing the σ -algebras $T^k \psi^{-1} \mathcal{G}$ for all $k \geq 0$.

Example. Let T on $(\{0, 1\}^{\mathbb{Z}}, \mathcal{F}, \mu)$ be the two-sided Bernoulli shift, and S on $(\{0, 1\}^{\mathbb{N} \cup \{0\}}, \mathcal{G}, \nu)$ be the one-sided Bernoulli shift, both spaces are endowed with the uniform product measure. Notice that T is invertible, while S is not. Set $X = \{0, 1\}^{\mathbb{Z}}$, $Y = \{0, 1\}^{\mathbb{N} \cup \{0\}}$, and define $\psi : X \rightarrow Y$ by

$$\psi(\dots, x_{-1}, x_0, x_1, \dots) = (x_0, x_1, \dots).$$

Then, ψ is a factor map. We claim that

$$\bigvee_{k=0}^{\infty} T^k \psi^{-1} \mathcal{G} = \mathcal{F}.$$

To prove this, we show that $\bigvee_{k=0}^{\infty} T^k \psi^{-1} \mathcal{G}$ contains all cylinders generating \mathcal{F} .

Let $\Delta = \{x \in X : x_{-k} = a_{-k}, \dots, x_{\ell} = a_{\ell}\}$ be an arbitrary cylinder in \mathcal{F} , and let $D = \{y \in Y : y_0 = a_{-k}, \dots, y_{k+\ell} = a_{\ell}\}$ which is a cylinder in \mathcal{G} . Then,

$$\psi^{-1} D = \{x \in X : x_0 = a_{-k}, \dots, x_{k+\ell} = a_{\ell}\} \quad \text{and} \quad T^k \psi^{-1} D = \Delta.$$

This shows that

$$\bigvee_{k=0}^{\infty} T^k \psi^{-1} \mathcal{G} = \mathcal{F}.$$

Thus, T is the natural extension of S .

Exercise 3.3.1 Consider $([0, 1], \mathcal{B}, \lambda)$, where \mathcal{B} is the Lebesgue σ -algebra, and λ is Lebesgue measure. Let $T : [0, 1) \rightarrow [0, 1)$ be defined by

$$Tx = \begin{cases} n(n+1)x - n & \text{if } x \in \left[\frac{1}{n+1}, \frac{1}{n}\right) \\ 0 & \text{if } x = 0. \end{cases}$$

Define $a_1 : [0, 1) \rightarrow [2, \infty]$ by

$$a_1 = a_1(x) = \begin{cases} n+1 & \text{if } x \in \left[\frac{1}{n+1}, \frac{1}{n}\right), n \geq 1 \\ \infty & \text{if } x = 0. \end{cases}$$

For $n \geq 1$, let $a_n = a_n(x) = a_1(T^{n-1}x)$.

- (a) Show that T is measure preserving with respect to Lebesgue measure λ .
- (b) Show that for λ a.e. x there exists a sequence a_1, a_2, \dots of positive integers such that $a_i \geq 2$ for all $i \geq 1$, and

$$x = \frac{1}{a_1} + \frac{1}{a_1(a_1-1)a_2} + \dots + \frac{1}{a_1(a_1-1) \cdots a_{k-1}(a_{k-1}-1)a_k} + \dots$$

- (c) Consider the dynamical system (X, \mathcal{F}, μ, S) , where $X = \{2, 3, \dots\}^{\mathbb{N}}$, \mathcal{F} the σ -algebra generated by the cylinder sets, S the left shift on X , and μ the product measure with $\mu(\{x : x_1 = j\}) = \frac{1}{j(j-1)}$. Show that $([0, 1), \mathcal{B}, \lambda, T)$ and (X, \mathcal{F}, μ, S) are isomorphic.
- (d) Show that T is strongly mixing.
- (e) Consider the product space $([0, 1) \times [0, 1), \mathcal{B} \times \mathcal{B}, \lambda \times \lambda)$. Define the transformation $\mathcal{T} : [0, 1) \times [0, 1) \rightarrow [0, 1) \times [0, 1)$ by

$$\mathcal{T}(x, y) = \begin{cases} (Tx, \frac{y+n}{n(n+1)}) & \text{if } x \in [\frac{1}{n+1}, \frac{1}{n}) \\ (0, 0) & \text{if } x = 0. \end{cases}$$

Show that $([0, 1) \times [0, 1), \mathcal{B} \times \mathcal{B}, \lambda \times \lambda, \mathcal{T})$ is a natural extension of $([0, 1), \mathcal{B}, \lambda, T)$.

Exercise 3.3.2

Consider $([0, 1), \mathcal{B}, \lambda)$, where \mathcal{B} is the Lebesgue σ -algebra, and λ is Lebesgue measure. Let $\beta > 1$ be a real number satisfying $\beta^3 = \beta^2 + \beta + 1$, and consider the β -transformation $T_\beta : [0, 1) \rightarrow [0, 1)$ given by $T_\beta x = \beta x \pmod{1}$. Define a measure ν on \mathcal{B} by

$$\nu(A) = \int_A h(x) dx,$$

where

$$h(x) = \begin{cases} \frac{1}{\frac{1}{\beta} + \frac{2}{\beta^2} + \frac{3}{\beta^3}} \left(1 + \frac{1}{\beta} + \frac{1}{\beta^2}\right) & \text{if } x \in [0, 1/\beta) \\ \frac{1}{\frac{1}{\beta} + \frac{2}{\beta^2} + \frac{3}{\beta^3}} \left(1 + \frac{1}{\beta}\right) & \text{if } x \in [1/\beta, 1/\beta + 1/\beta^2) \\ \frac{1}{\frac{1}{\beta} + \frac{2}{\beta^2} + \frac{3}{\beta^3}} \cdot 1 & \text{if } x \in [1/\beta + 1/\beta^2, 1) \end{cases}$$

- (a) Show that T_β is measure preserving with respect to ν .
- (b) Let

$$X = \left([0, \frac{1}{\beta}) \times [0, 1)\right) \times \left([\frac{1}{\beta}, \frac{1}{\beta} + \frac{1}{\beta^2}) \times [0, \frac{1}{\beta} + \frac{1}{\beta^2})\right) \times \left([\frac{1}{\beta} + \frac{1}{\beta^2}, 1) \times [0, \frac{1}{\beta})\right).$$

Let \mathcal{C} be the restriction of the two dimensional Lebesgue σ -algebra on X , and μ the normalized (two dimensional) Lebesgue measure on X . Define on X the transformation \mathcal{T}_β as follows :

$$\mathcal{T}_\beta(x, y) := \left(T_\beta x, \frac{1}{\beta}(\lfloor \beta x \rfloor + y) \right) \quad \text{for } (x, y) \in X.$$

- (i) Show that \mathcal{T}_β is measurable and measure preserving with respect to μ . Prove also that \mathcal{T}_β is one-to-one and onto μ a.e.
- (ii) Show that \mathcal{T}_β is the natural extension of T_β .

Chapter 4

Continued Fractions

4.1 Introduction and Basic Properties

Recall from Chapter 1 that any real number $x \in [0, 1)$ can be written as a continued fraction of the form

$$x = \frac{1}{a_1 + \frac{1}{a_2 + \frac{1}{a_3 + \ddots}}}$$

Such an expansion is generated by the map T defined on $[0, 1)$ by

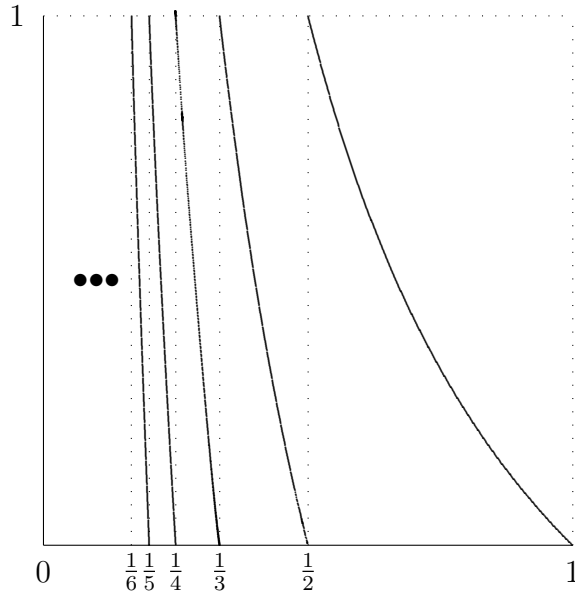
$$Tx = \frac{1}{x} - \left\lfloor \frac{1}{x} \right\rfloor;$$

see Figure 4.1. The digits a_n are defined by $a_n = a_n(x) = a_1(T^{n-1}x)$, where

$$a_1 = a_1(x) = \begin{cases} 1 & \text{if } x \in (\frac{1}{2}, 1) \\ n & \text{if } x \in (\frac{1}{n+1}, \frac{1}{n}], n \geq 2. \end{cases}$$

After n iteration of the map T one gets

$$x = \frac{1}{a_1 + Tx} = \dots = \frac{1}{a_1 + \frac{1}{a_2 + \dots + \frac{1}{a_n + T^n x}}}$$

Figure 4.1: The continued fraction map T .

Let

$$\frac{p_n}{q_n} = \frac{1}{a_1 + \frac{1}{a_2 + \cdots + \frac{1}{a_n}}},$$

we will now give some basic properties of this sequence and show that $x = \lim_{n \rightarrow \infty} \frac{p_n}{q_n}$. This done by studying elementary properties of matrices associated with the continued fraction expansion.

Let $A \in \text{SL}_2(\mathbb{Z})$, that is

$$A = \begin{bmatrix} r & p \\ s & q \end{bmatrix},$$

where $r, s, p, q \in \mathbb{Z}$ and $\det A = rq - ps \in \{\pm 1\}$. Now define the *Möbius* (or: *fractional linear*) *transformation* $A: \mathbb{C}^* \rightarrow \mathbb{C}^*$ by

$$A(z) = \begin{bmatrix} r & p \\ s & q \end{bmatrix} (z) = \frac{rz + p}{sz + q}.$$

Let a_1, a_2, \dots be the sequence of partial quotients of x . Put

$$A_n := \begin{bmatrix} 0 & 1 \\ 1 & a_n \end{bmatrix}, \quad n \geq 1 \tag{4.1}$$

and

$$M_n := A_1 A_2 \cdots A_n, \quad n \geq 1.$$

Writing

$$M_n := \begin{bmatrix} r_n & p_n \\ s_n & q_n \end{bmatrix}, \quad n \geq 1,$$

it follows from $M_n = M_{n-1}A_n$, $n \geq 2$, that

$$\begin{bmatrix} r_n & p_n \\ s_n & q_n \end{bmatrix} = \begin{bmatrix} r_{n-1} & p_{n-1} \\ s_{n-1} & q_{n-1} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & a_n \end{bmatrix},$$

yielding the recurrence relations

$$p_{-1} := 1; \quad p_0 := a_0 = 0; \quad p_n = a_n p_{n-1} + p_{n-2}, \quad n \geq 1, \quad (4.2)$$

$$q_{-1} := 0; \quad q_0 := 1; \quad q_n = a_n q_{n-1} + q_{n-2}, \quad n \geq 1.$$

Furthermore, $p_n(x) = q_{n-1}(Tx)$ for all $n \geq 0$, and $x \in (0, 1)$.

Now

$$\omega_n = M_n(0) = \frac{p_n}{q_n}$$

and from $\det M_n = (-1)^n$ it follows, that

$$p_{n-1}q_n - p_nq_{n-1} = (-1)^n, \quad n \geq 1, \quad (4.3)$$

hence

$$\gcd(p_n, q_n) = 1, \quad n \geq 1,$$

Setting

$$A_n^* := \begin{bmatrix} 0 & 1 \\ 1 & a_n + T^n x \end{bmatrix},$$

it follows from

$$x = M_{n-1}A_n^*(0) = [0; a_1, \dots, a_{n-1}, a_n + T^n x],$$

that

$$x = \frac{p_n + p_{n-1}T^n x}{q_n + q_{n-1}T^n x}, \quad (4.4)$$

i.e., $x = M_n(T^n x)$. From this and (4.3) one at once has that

$$x - \frac{p_n}{q_n} = \frac{(-1)^n T^n x}{q_n(q_n + q_{n-1}T^n x)}. \quad (4.5)$$

In fact, (4.5) yields information about the quality of approximation of the rational number $\omega_n = p_n/q_n$ to the irrational number x . Since $T^n x \in [0, 1)$, it at once follows that

$$\left| x - \frac{p_n}{q_n} \right| < \frac{1}{q_n^2}, \quad n \geq 0. \quad (4.6)$$

From $1/T^n x = a_{n+1} + T^{n+1}x$ one even has

$$\frac{1}{2q_n q_{n+1}} < \left| x - \frac{p_n}{q_n} \right| < \frac{1}{q_n q_{n+1}}, \quad n \geq 1.$$

Notice that the recurrence relations (4.2) yield that

$$\omega_n - \omega_{n-1} = \frac{(-1)^{n+1}}{q_{n-1}q_n}, \quad n \geq 1. \quad (4.7)$$

From this and (4.5) one sees, that

$$0 = \omega_0 < \omega_2 < \omega_4 < \cdots < \omega_3 < \omega_1 < 1. \quad (4.8)$$

In view of proposition 1 the following questions arise naturally: “Given a sequence of positive integers $(a_n)_{n \geq 1}$, does $\lim_{n \rightarrow \infty} \omega_n$ exist? Moreover, in case the limit exists and equals x , do we have that $x = [0; a_1, \dots, a_n, \dots]$?”

We have the following proposition.

Proposition 4.1.1 *Let $(a_n)_{n \geq 1}$ be a sequence of positive integers, and let the sequence of rational numbers $(\omega_n)_{n \geq 1}$ be given by*

$$\omega_n := [0; a_1, \dots, a_n], \quad n \geq 1.$$

Then there exists an irrational number x for which

$$\lim_{n \rightarrow \infty} \omega_n = x$$

and we moreover have that $x = [0; a_1, a_2, \dots, a_n, \dots]$.

Proof. Writing $\omega_n = p_n/q_n$, $n \geq 1$, $\omega_0 := 0$, where

$$\begin{bmatrix} p_{n-1} & p_n \\ q_{n-1} & q_n \end{bmatrix} = A_1 \cdots A_n$$

and where A_i is defined as in (4.1), one has from (4.7), (4.8) and $\omega_0 := 0$ that

$$\omega_n = \sum_{k=1}^n \frac{(-1)^{k+1}}{q_{k-1}q_k},$$

hence Leibniz' theorem yields that $\lim_{n \rightarrow \infty} \omega_n$ exists and equals, say, x . In order to show that the sequence of positive integers $(a_n)_{n \geq 1}$ determines a unique $x \in \mathbb{R}$ we have to show that $a_n = a_n(x)$ for $n \geq 1$, i.e. that $(a_n)_{n \geq 1}$ is the sequence of partial quotients of x . Since

$$\omega_n = [0; a_1, a_2, \dots, a_n] = \frac{1}{a_1 + [0; a_2, a_3, \dots, a_n]}$$

it is sufficient to show that

$$\left[\frac{1}{x} \right] = a_1.$$

However,

$$\omega_n = \frac{1}{a_1 + \omega_n^*},$$

where $\omega_n^* = [0; a_2, a_3, \dots, a_n]$. Hence taking limits $n \rightarrow \infty$ yields

$$x = \frac{1}{a_1 + x^*},$$

here $x^* = \lim_{n \rightarrow \infty} \omega_n^*$. From $0 < \omega_2^* < x^* < \omega_3^* < 1$, see also (4.8), and from $1/x = a_1 + x^*$ it now follows that $[1/x] = a_1$. \square

Exercise 4.1.1 Let

$$\Delta(a_1, \dots, a_n) := \{x \in [0, 1) : a_1(x) = a_1, \dots, a_n(x) = a_n\},$$

where $a_j \in \mathbb{N}$ for each $1 \leq j \leq n$. Show that $\Delta(a_1, a_2, \dots, a_n)$ is an interval in $[0, 1)$ with endpoints

$$\frac{p_k}{q_k} \quad \text{and} \quad \frac{p_k + p_{k-1}}{q_k + q_{k-1}},$$

where

$$\frac{p_n}{q_n} = \frac{1}{a_1 + \frac{1}{a_2 + \dots + \frac{1}{a_n}}}.$$

Conclude that

$$\lambda(\Delta(a_1, a_2, \dots, a_k)) = \frac{1}{q_k(q_k + q_{k-1})},$$

where λ is Lebesgue measure on $[0, 1)$. We call such an interval, a *fundamental interval of order n* .

Exercise 4.1.2 Let $0 \leq a < b \leq 1$. Show that $\{x : a \leq T^n x < b\} \cap \Delta_n(a_1, \dots, a_n)$ equals

$$\left[\frac{p_{n-1}a + p_n}{q_{n-1}a + q_n}, \frac{p_{n-1}b + p_n}{q_{n-1}b + q_n} \right)$$

when n is even, and equals

$$\left(\frac{p_{n-1}b + p_n}{q_{n-1}b + q_n}, \frac{p_{n-1}a + p_n}{q_{n-1}a + q_n} \right]$$

for n odd. Conclude that

$$\lambda(T^{-n}[a, b] \cap \Delta_n) = \lambda([a, b))\lambda(\Delta_n) \frac{q_n(q_{n-1} + q_n)}{(q_{n-1}b + q_n)(q_{n-1}a + q_n)},$$

where λ is Lebesgue measure on $[0, 1)$.

4.2 Ergodic Properties of Continued Fraction Map

In Exercise 1.3.4 we saw that the continued fraction map is measure preserving with respect to the Gauss measure μ given by

$$\mu(B) = \int_B \frac{1}{\log 2} \frac{1}{1+x} dx.$$

We will now show that T is ergodic with respect to μ .

Theorem 4.2.1 *Let T be the continued fraction map, and μ Gauss measure on $[0, 1)$. Then, T is ergodic with respect to μ .*

Proof. Let $[a, b)$ be any interval in $[0, 1)$, and $\Delta_n = \Delta_n(a_1, \dots, a_n)$ a fundamental interval of order n . From Exercise 4.1.2, we have

$$\lambda(T^{-n}[a, b) \cap \Delta_n) = \lambda([a, b))\lambda(\Delta_n) \frac{q_n(q_{n-1} + q_n)}{(q_{n-1}b + q_n)(q_{n-1}a + q_n)}.$$

Since

$$\frac{1}{2} < \frac{q_n}{q_{n-1} + q_n} < \frac{q_n(q_{n-1} + q_n)}{(q_{n-1}b + q_n)(q_{n-1}a + q_n)} < \frac{q_n(q_{n-1} + q_n)}{q_n^2} < 2.$$

Therefore we find for every interval I , that

$$\frac{1}{2}\lambda(I)\lambda(\Delta_n) < \lambda(T^{-n}I \cap \Delta_n) < 2\lambda(I)\lambda(\Delta_n).$$

Let A be a finite disjoint union of such intervals I . Since Lebesgue measure is additive one has

$$\frac{1}{2}\lambda(A)\lambda(\Delta_n) \leq \lambda(T^{-n}A \cap \Delta_n) \leq 2\lambda(A)\lambda(\Delta_n). \quad (4.9)$$

The collection of finite disjoint unions of such intervals generates the Borel σ -algebra. It follows that (4.9) holds for any Borel set A . Since

$$\frac{1}{2\log 2}\lambda(A) \leq \mu(A) \leq \frac{1}{\log 2}\lambda(A), \quad (4.10)$$

then by (4.9) and (4.10) one has

$$\mu(T^{-n}A \cap \Delta_n) \geq \frac{\log 2}{4}\mu(A)\mu(\Delta_n). \quad (4.11)$$

Now let \mathcal{C} be the collection of all fundamental intervals Δ_n . Since the set of all endpoints of these fundamental intervals is the set of all rationals in $[0, 1)$, it follows that condition (a) of Knopp's Lemma is satisfied. Now suppose that B is invariant with respect to T and $\mu(B) > 0$. Then it follows from (4.11) that for every fundamental interval Δ_n

$$\mu(B \cap \Delta_n) \geq \frac{\log 2}{4}\mu(B)\mu(\Delta_n).$$

So condition (b) from Knopp's Lemma is satisfied with $\gamma = \frac{\log 2}{4}\mu(B)$; thus $\mu(B) = 1$; i.e. T is ergodic. \square

We now use the ergodic Theorem to give simple proofs of old and famous results of Paul Lévy; see [Le].

Proposition 4.2.1 (Paul Lévy, 1929) *For almost all $x \in [0, 1)$ one has*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log q_n = \frac{\pi^2}{12 \log 2}, \quad (4.12)$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log(\lambda(\Delta_n)) = \frac{-\pi^2}{6 \log 2}, \quad \text{and} \quad (4.13)$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log \left| x - \frac{p_n}{q_n} \right| = \frac{-\pi^2}{6 \log 2}. \quad (4.14)$$

Proof. By the recurrence relations (4.2), for any irrational $x \in [0, 1)$ one has

$$\begin{aligned} \frac{1}{q_n(x)} &= \frac{1}{q_n(x)} \frac{p_n(x)}{q_{n-1}(Tx)} \frac{p_{n-1}(Tx)}{q_{n-2}(T^2x)} \cdots \frac{p_2(T^{n-2}x)}{q_1(T^{n-1}x)} \\ &= \frac{p_n(x)}{q_n(x)} \frac{p_{n-1}(Tx)}{q_{n-1}(Tx)} \cdots \frac{p_1(T^{n-1}x)}{q_1(T^{n-1}x)}. \end{aligned}$$

Taking logarithms yields

$$-\log q_n(x) = \log \frac{p_n(x)}{q_n(x)} + \log \frac{p_{n-1}(Tx)}{q_{n-1}(Tx)} + \cdots + \log \frac{p_1(T^{n-1}x)}{q_1(T^{n-1}x)}. \quad (4.15)$$

For any $k \in \mathbb{N}$, and any irrational $x \in [0, 1)$, $\frac{p_k(x)}{q_k(x)}$ is a rational number close to x . Therefore we compare the right-hand side of (4.15) with

$$\log x + \log Tx + \log T^2x + \cdots + \log(T^{n-1}x).$$

We have

$$-\log q_n(x) = \log x + \log Tx + \log T^2x + \cdots + \log(T^{n-1}x) + R(n, x).$$

In order to estimate the error term $R(n, x)$, we recall from Exercise 4.1.1 that x lies in the interval Δ_n , which has endpoints p_n/q_n and $(p_n + p_{n-1})/(q_n + q_{n-1})$. Therefore, in case n is even, one has

$$0 < \log x - \log \frac{p_n}{q_n} = \left(x - \frac{p_n}{q_n}\right) \frac{1}{\xi} \leq \frac{1}{q_n(q_{n-1} + q_n)} \frac{1}{p_n/q_n} < \frac{1}{q_n},$$

where $\xi \in (p_n/q_n, x)$ is given by the mean value theorem. Let $\mathcal{F}_1, \mathcal{F}_2, \dots$ be the sequence of Fibonacci 1, 1, 2, 3, 5, \dots (these are the q_i 's of the small

golden ratio $g = 1/G$). It follows from the recurrence relation for the q_i 's that $q_n(x) \geq \mathcal{F}_n$. A similar argument shows that

$$\frac{1}{q_n} < \log x - \log \frac{p_n}{q_n},$$

in case n is odd. Thus

$$|R(n, x)| \leq \frac{1}{\mathcal{F}_n} + \frac{1}{\mathcal{F}_{n-1}} + \cdots + \frac{1}{\mathcal{F}_1},$$

and since we have

$$\mathcal{F}_n = \frac{G^n + (-1)^{n+1}g^n}{\sqrt{5}}$$

it follows that $\mathcal{F}_n \sim \frac{1}{\sqrt{5}}G^n$, $n \rightarrow \infty$. Thus $\frac{1}{\mathcal{F}_n} + \frac{1}{\mathcal{F}_{n-1}} + \cdots + \frac{1}{\mathcal{F}_1}$ is the n th partial sum of a convergent series, and therefore

$$|R(n, x)| \leq \frac{1}{\mathcal{F}_n} + \cdots + \frac{1}{\mathcal{F}_1} \leq \sum_{n=1}^{\infty} \frac{1}{\mathcal{F}_n} := \mathcal{C}.$$

Hence for each x for which

$$\lim_{n \rightarrow \infty} \frac{1}{n} (\log x + \log Tx + \log T^2x + \cdots + \log(T^{n-1}x))$$

exists,

$$- \lim_{n \rightarrow \infty} \frac{1}{n} \log q_n(x)$$

exists too, and these limits are equal.

Now $\lim_{n \rightarrow \infty} \frac{1}{n} (\log x + \log Tx + \log T^2x + \cdots + \log(T^{n-1}x))$ is ideally suited for the Pointwise Ergodic Theorem; we only need to check that the conditions of the Ergodic Theorem are satisfied and to calculate the integral. This is left as an exercise for the reader. This proves (4.12).

From Exercise 4.1.1

$$\lambda(\Delta_n(a_1, \dots, a_n)) = \frac{1}{q_n(q_n + q_{n-1})};$$

thus

$$-\log 2 - 2 \log q_n < \log \lambda(\Delta_n) < -2 \log q_n.$$

Now apply (4.12) to obtain (4.13). Finally (4.14) follows from (4.12) and

$$\frac{1}{2q_n q_{n+1}} < \left| x - \frac{p_n}{q_n} \right| < \frac{1}{q_n q_{n+1}}, \quad n \geq 1.$$

Exercise 4.2.1 Consider $([0, 1], \mathcal{B})$, where \mathcal{B} is the Lebesgue σ -algebra. Let $T : [0, 1) \rightarrow [0, 1)$ be the *Continued fraction* transformation, i.e., $T0 = 0$ and for $x \neq 0$

$$Tx = \frac{1}{x} - \left\lfloor \frac{1}{x} \right\rfloor.$$

For each $x \in [0, 1)$ consider the sequence of digits of x defined by $a_n(x) = a_n = \left\lfloor \frac{1}{T^{n-1}x} \right\rfloor$. Let λ denote the normalized Lebesgue measure on $[0, 1)$.

(a) Show that

$$\lim_{n \rightarrow \infty} (a_1 a_2 \dots a_n)^{1/n} = \prod_{k=1}^{\infty} \left(1 + \frac{1}{k(k+2)} \right)^{\frac{\log k}{\log 2}}$$

λ a.e.

(b) Show that $\lim_{n \rightarrow \infty} \frac{a_1 + a_2 + \dots + a_n}{n} = \infty$ λ a.e.

4.3 Natural Extension and the Doeblin-Lenstra Conjecture

A planar and a very useful version of a natural extension of the Continued fraction map was given by Ito-Nakada-Tanaka.

Theorem 4.3.1 (Ito, Nakada, Tanaka, 1977; Nakada, 1981) *Let $\bar{\Omega} = [0, 1) \times [0, 1]$, $\bar{\mathcal{B}}$ be the collection of Borel sets of $\bar{\Omega}$. Define the two-dimensional Gauss-measure $\bar{\mu}$ on $(\bar{\Omega}, \bar{\mathcal{B}})$ by*

$$\bar{\mu}(E) = \frac{1}{\log 2} \iint_E \frac{dx dy}{(1+xy)^2}, \quad E \in \bar{\mathcal{B}}.$$

Finally, let the two-dimensional RCF-operator $\mathcal{T} : \bar{\Omega} \rightarrow \bar{\Omega}$ for $(x, y) \in \bar{\Omega}$ be defined by

$$\mathcal{T}(x, y) = \left(T(x), \frac{1}{\lfloor \frac{1}{x} \rfloor + y} \right), \quad x \neq 0, \quad \mathcal{T}(0, y) = (0, y). \quad (4.16)$$

Then $(\bar{\Omega}, \bar{\mathcal{B}}, \bar{\mu}, \mathcal{T})$ is the natural extension of $([0, 1], \mathcal{B}, \mu, T)$. Furthermore, it is ergodic.

Proof. We leave the proof that \mathcal{T} is the natural extension of T as an exercise below. We just show that \mathcal{T} is measure preserving with respect to $\bar{\mu}$. The map \mathcal{T} is a bijective mapping from $\bar{\Omega}$ to $\bar{\Omega}$. For $(x, y) \in \bar{\Omega}$ let $(\xi, \eta) \in \bar{\Omega}$ be such, that $(\xi, \eta) = \mathcal{T}(x, y)$. Then

$$\xi = \frac{1}{x} - c \Leftrightarrow x = \frac{1}{c + \xi},$$

and

$$\eta = \frac{1}{c + y} \Leftrightarrow y = \frac{1}{\eta} - c.$$

Hence the above coordinate transformation has Jacobian J , which satisfies

$$J = \begin{vmatrix} \frac{\partial x}{\partial \xi} & \frac{\partial x}{\partial \eta} \\ \frac{\partial y}{\partial \xi} & \frac{\partial y}{\partial \eta} \end{vmatrix} = \begin{vmatrix} \frac{-1}{(c+\xi)^2} & 0 \\ 0 & \frac{-1}{\eta^2} \end{vmatrix} = \frac{1}{(c+\xi)^2} \frac{1}{\eta^2},$$

and therefore we find

$$\begin{aligned} \bar{\mu}(A) &= \frac{1}{\log 2} \iint_A \frac{dx dy}{(1 + xy)^2} \\ &= \frac{1}{\log 2} \iint_{\mathcal{T}A} \frac{d\xi d\eta}{\left(1 + \frac{1}{c+\xi} \left(\frac{1}{\eta} - c\right)\right)^2} \frac{1}{(c + \xi)^2 \eta^2} \\ &= \frac{1}{\log 2} \iint_{\mathcal{T}A} \frac{d\xi d\eta}{(1 + \xi\eta)^2} \\ &= \bar{\mu}(\mathcal{T}A). \end{aligned}$$

□

Exercise 4.3.1 Prove Theorem 4.3.1.

We define for every real number $x \in [0, 1)$ and every $n \geq 0$ the so-called *approximation coefficients* $\Theta_n(x)$ by

$$\Theta_n(x) = q_n^2 \left| x - \frac{p_n}{q_n} \right|. \quad (4.17)$$

It follows from (4.6) that $\Theta_n(x) < 1$ for all irrational $x \in [0, 1)$ and all $n \geq 0$. In 1981, H.W. Lenstra conjectured that for almost all x the limit

$$\lim_{n \rightarrow \infty} \frac{1}{n} \#\{j; 1 \leq j \leq n, \Theta_j(x) \leq z\}, \text{ where } 0 \leq z \leq 1,$$

exists, and equals the distribution function $F(z)$, given by

$$F(z) \begin{cases} \frac{z}{\log 2} & 0 \leq z \leq \frac{1}{2} \\ \frac{1}{\log 2}(1 - z + \log 2z) & \frac{1}{2} \leq z \leq 1, \end{cases} \quad (4.18)$$

where the $\Theta_n(x)$ s are the approximation coefficients as defined in (4.17).

In other words: for almost all x the sequence $(\Theta_n(x))_{n \geq 1}$ has limiting distribution F .

An immediate corollary of this conjecture is that for almost all x

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^{n-1} \Theta_j(x) = \frac{1}{4 \log 2} = 0.360673 \dots$$

A first attempt at Lenstra's conjecture was made by D.S. Knuth ([Knu]), who obtained the following theorem

Theorem 4.3.2 (Knuth, 1984) *Let $K_n(z) = \{x \in [0, 1) \setminus \mathbb{Q}; \Theta_n \leq z\}$ for $0 \leq z \leq 1$, then*

$$\lambda(K_n(z)) = F(z) + \mathcal{O}(g^n),$$

where F is defined as in (4.18).

See also [DK] for a generalization of this result.

By definition of the approximation coefficients we have

$$\Theta_n = q_n^2 \left| x - \frac{p_n}{q_n} \right| = \frac{T^n(x)}{1 + T^n(x) \frac{q_{n-1}}{q_n}}.$$

By the recurrence-relations (4.2) we have

$$\frac{q_{n-1}}{q_n} = \frac{q_{n-1}}{a_n q_{n-1} + q_{n-2}} = \frac{1}{a_n + \frac{q_{n-2}}{q_{n-1}}},$$

so setting $V_n = V_n(x) = q_{n-1}/q_n$ yields for $n \geq 1$ that

$$V_n = \frac{1}{a_n + V_{n-1}} = \dots = \frac{1}{a_n + \frac{1}{a_{n-1} + \frac{1}{a_{n-1} + \dots + \frac{1}{a_1}}}} = [0; a_n, a_{n-1}, \dots, a_1].$$

Exercise 4.3.2 (i) Show that for any $x \in [0, 1)$ one has $\mathcal{T}(x, 0) = (T_n, V_n)$.

(ii) Show that for any $x \in [0, 1)$ one has

$$\lim_{n \rightarrow \infty} (\mathcal{T}^n(x, 0) - \mathcal{T}^n(x, y)) = 0,$$

uniformly in y .

From the above, we can interpret V_n as “the past of x at time n ” (in the same way as $T^n(x)$ is the “future of x at time n ”). An immediate consequence of this and (4.5) is that

$$\Theta_n = \Theta_n(x) = \frac{T^n(x)}{1 + T^n(x)V_n}, \quad n \geq 0. \quad (4.19)$$

Exercise 4.3.3 show that

$$\Theta_{n-1} = \Theta_{n-1}(x) = \frac{V_n}{1 + T^n(x)V_n}, \quad n \geq 1. \quad (4.20)$$

An important consequence of this observation and Theorem 4.3.1 is the following result.

Theorem 4.3.3 (Jager, 1986) *For almost all $x \in [0, 1)$ the two-dimensional sequence $(T^n(x), V_n(x))_{n \geq 1}$ is distributed over $\bar{\Omega}$ according to the density function d , where*

$$d(x, y) = \frac{1}{\log 2} \frac{1}{(1 + xy)^2}.$$

Proof. Denote by E the subset of numbers $x \in \Omega$ for which the sequence $(T_n, V_n)_{n \geq 0}$ is **not** distributed according to the density function d . Since the sequence $(\mathcal{T}^n(x, 0) - \mathcal{T}^n(x, y))_{n \geq 0}$ is a null-sequence, it follows that for every pair $(x, y) \in \mathcal{E}$, where $\mathcal{E} := E \times [0, 1]$, the sequence $\mathcal{T}^n(x, y)_{n \geq 0}$ is **not** distributed according to the density function d . Now if E had, as a one-dimensional set, positive Lebesgue measure, so would \mathcal{E} as a two-dimensional set. But this would be in conflict with Theorem 4.3.1. \square

Lenstra's conjecture now follows directly from this theorem by easy calculations.

Proof of Lenstra's conjecture. Let $A_z = \{(x, y) \in \bar{\Omega} : \frac{x}{1+xy} \leq z\}$. Then, $\bar{\mu}(A_z) = F(z)$, and $\Theta_j(x) \leq z \Leftrightarrow \mathcal{T}^j(x, 0) \in A_z$. Furthermore,

$$\frac{1}{n} \#\{j; 1 \leq j \leq n, \Theta_j(x) \leq z\} = \frac{1}{n} \sum_{j=1}^n \mathbf{1}_{A_z}(\mathcal{T}^j(x, 0)).$$

Taking limits and using the above theorem we get the required result.

Chapter 5

Entropy

5.1 Randomness and Information

Given a measure preserving transformation T on a probability space (X, \mathcal{F}, μ) , we want to define a nonnegative quantity $h(T)$ which measures the average uncertainty about where T moves the points of X . That is, the value of $h(T)$ reflects the amount of ‘randomness’ generated by T . We want to define $h(T)$ in such a way, that (i) the amount of information gained by an application of T is proportional to the amount of uncertainty removed, and (ii) that $h(T)$ is isomorphism invariant, so that isomorphic transformations have equal entropy.

The connection between entropy (that is randomness, uncertainty) and the transmission of information was first studied by Claude Shannon in 1948. As a motivation let us look at the following simple example. Consider a source (for example a ticker-tape) that produces a string of symbols $\cdots x_{-1}x_0x_1\cdots$ from the alphabet $\{a_1, a_2, \dots, a_n\}$. Suppose that the probability of receiving symbol a_i at any given time is p_i , and that each symbol is transmitted independently of what has been transmitted earlier. Of course we must have here that each $p_i \geq 0$ and that $\sum_i p_i = 1$. In ergodic theory we view this process as the dynamical system $(X, \mathcal{F}, \mathcal{B}, \mu, T)$, where $X = \{a_1, a_2, \dots, a_n\}^{\mathbb{N}}$, \mathcal{B} the σ -algebra generated by cylinder sets of the form

$$\Delta_n(a_{i_1}, a_{i_2}, \dots, a_{i_n}) := \{x \in X : x_{i_1} = a_{i_1}, \dots, x_{i_n} = a_{i_n}\}$$

μ the product measure assigning to each coordinate probability p_i of seeing

the symbol a_i , and T the left shift. We define the entropy of this system by

$$H(p_1, \dots, p_n) = h(T) := - \sum_{i=1}^n p_i \log_2 p_i. \quad (5.1)$$

If we define $\log p_i$ as the amount of uncertainty in transmitting the symbol a_i , then H is the average amount of information gained (or uncertainty removed) per symbol (notice that H is in fact an expected value). To see why this is an appropriate definition, notice that if the source is degenerate, that is, $p_i = 1$ for some i (i.e., the source only transmits the symbol a_i), then $H = 0$. In this case we indeed have no randomness. Another reason to see why this definition is appropriate, is that H is maximal if $p_i = \frac{1}{n}$ for all i , and this agrees with the fact that the source is most random when all the symbols are equiprobable. To see this maximum, consider the function $f : [0, 1] \rightarrow \mathbb{R}_+$ defined by

$$f(t) = \begin{cases} 0 & \text{if } t = 0, \\ -t \log_2 t & \text{if } 0 < t \leq 1. \end{cases}$$

Then f is continuous and concave downward, and Jensen's Inequality implies that for any p_1, \dots, p_n with $p_i \geq 0$ and $p_1 + \dots + p_n = 1$,

$$\frac{1}{n} H(p_1, \dots, p_n) = \frac{1}{n} \sum_{i=1}^n f(p_i) \leq f\left(\frac{1}{n} \sum_{i=1}^n p_i\right) = f\left(\frac{1}{n}\right) = \frac{1}{n} \log_2 n,$$

so $H(p_1, \dots, p_n) \leq \log_2 n$ for all probability vectors (p_1, \dots, p_n) . But

$$H\left(\frac{1}{n}, \dots, \frac{1}{n}\right) = \log_2 n,$$

so the maximum value is attained at $(\frac{1}{n}, \dots, \frac{1}{n})$.

5.2 Definitions and Properties

So far H is defined as the average information per symbol. The above definition can be extended to define the information transmitted by the occurrence of an event E as $-\log_2 P(E)$. This definition has the property that the information transmitted by $E \cap F$ for independent events E and F is the sum of the information transmitted by each one individually, i.e.,

$$-\log_2 P(E \cap F) = -\log_2 P(E) - \log_2 P(F).$$

The only function with this property is the logarithm function to any base. We choose base 2 because information is usually measured in bits.

In the above example of the ticker-tape the symbols were transmitted independently. In general, the symbol generated might depend on what has been received before. In fact these dependencies are often ‘built-in’ to be able to check the transmitted sequence of symbols on errors (think here of the Morse sequence, sequences on compact discs etc.). Such dependencies must be taken into consideration in the calculation of the average information per symbol. This can be achieved if one replaces the symbols a_i by blocks of symbols of particular size. More precisely, for every n , let \mathcal{C}_n be the collection of all possible n -blocks (or cylinder sets) of length n , and define

$$H_n := - \sum_{C \in \mathcal{C}_n} P(C) \log P(C).$$

Then $\frac{1}{n}H_n$ can be seen as the average information per symbol when a block of length n is transmitted. The entropy of the source is now defined by

$$h := \lim_{n \rightarrow \infty} \frac{H_n}{n}. \quad (5.2)$$

The existence of the limit in (5.2) follows from the fact that H_n is a *subadditive sequence*, i.e., $H_{n+m} \leq H_n + H_m$, and proposition (5.2.2) (see proposition (5.2.3) below).

Now replace the source by a measure preserving system (X, \mathcal{B}, μ, T) . How can one define the entropy of this system similar to the case of a source? The symbols $\{a_1, a_2, \dots, a_n\}$ can now be viewed as a partition $\mathcal{A} = \{A_1, A_2, \dots, A_n\}$ of X , so that X is the disjoint union (up to sets of measure zero) of A_1, A_2, \dots, A_n . The source can be seen as follows: with each point $x \in X$, we associate an infinite sequence $\dots x_{-1}, x_0, x_1, \dots$, where x_i is a_j if and only if $T^i x \in A_j$. We define the *entropy of the partition* α by

$$H(\alpha) = H_\mu(\alpha) := - \sum_{i=1}^n \mu(A_i) \log \mu(A_i).$$

Our aim is to define the entropy of the transformation T which is independent of the partition we choose. In fact $h(T)$ must be the maximal entropy over all possible finite partitions. But first we need few facts about partitions.

Exercise 5.2.1 Let $\alpha = \{A_1, \dots, A_n\}$ and $\beta = \{B_1, \dots, B_m\}$ be two partitions of X . Show that

$$T^{-1}\alpha := \{T^{-1}A_1, \dots, T^{-1}A_n\}$$

and

$$\alpha \vee \beta := \{A_i \cap B_j : A_i \in \alpha, B_j \in \beta\}$$

are both partitions of X .

The members of a partition are called the *atoms* of the partition. We say that the partition $\beta = \{B_1, \dots, B_m\}$ is a *refinement* of the partition $\alpha = \{A_1, \dots, A_n\}$, and write $\alpha \leq \beta$, if for every $1 \leq j \leq m$ there exists an $1 \leq i \leq n$ such that $B_j \subset A_i$ (up to sets of measure zero). The partition $\alpha \vee \beta$ is called the *common refinement* of α and β .

Exercise 5.2.2 Show that if β is a refinement of α , each atom of α is a finite (disjoint) union of atoms of β .

Given two partitions $\alpha = \{A_1, \dots, A_n\}$ and $\beta = \{B_1, \dots, B_m\}$ of X , we define the *conditional entropy of α given β* by

$$H(\alpha|\beta) := - \sum_{A \in \alpha} \sum_{B \in \beta} \log \left(\frac{\mu(A \cap B)}{\mu(B)} \right) \mu(A \cap B).$$

(Under the convention that $0 \log 0 := 0$.)

The above quantity $H(\alpha|\beta)$ is interpreted as the average uncertainty about which element of the partition α the point x will enter (under T) if we already know which element of β the point x will enter.

Proposition 5.2.1 Let α , β and γ be partitions of X . Then,

- (a) $H(T^{-1}\alpha) = H(\alpha)$;
- (b) $H(\alpha \vee \beta) = H(\alpha) + H(\beta|\alpha)$;
- (c) $H(\beta|\alpha) \leq H(\beta)$;
- (d) $H(\alpha \vee \beta) \leq H(\alpha) + H(\beta)$;
- (e) If $\alpha \leq \beta$, then $H(\alpha) \leq H(\beta)$;

$$(f) \quad H(\alpha \vee \beta | \gamma) = H(\alpha | \gamma) + H(\beta | \alpha \vee \gamma);$$

$$(g) \quad \text{If } \beta \leq \alpha, \text{ then } H(\gamma | \alpha) \leq H(\gamma | \beta);$$

$$(h) \quad \text{If } \beta \leq \alpha, \text{ then } H(\beta | \alpha) = 0.$$

(i) We call two partitions α and β independent if

$$\mu(A \cap B) = \mu(A)\mu(B) \quad \text{for all } A \in \alpha, B \in \beta.$$

If α and β are independent partitions, one has that

$$H(\alpha \vee \beta) = H(\alpha) + H(\beta).$$

Proof. We prove Properties (b) and (c), the rest are left as an exercise.

$$\begin{aligned} H(\alpha \vee \beta) &= - \sum_{A \in \alpha} \sum_{B \in \beta} \mu(A \cap B) \log \mu(A \cap B) \\ &= - \sum_{A \in \alpha} \sum_{B \in \beta} \mu(A \cap B) \log \frac{\mu(A \cap B)}{\mu(A)} \\ &\quad - \sum_{A \in \alpha} \sum_{B \in \beta} \mu(A \cap B) \log \mu(A) \\ &= H(\beta | \alpha) + H(\alpha). \end{aligned}$$

We now show that $H(\beta | \alpha) \leq H(\beta)$. Recall that the function $f(t) = -t \log t$ for $0 < t \leq 1$ is concave down. Thus,

$$\begin{aligned} H(\beta | \alpha) &= - \sum_{B \in \beta} \sum_{A \in \alpha} \mu(A \cap B) \log \frac{\mu(A \cap B)}{\mu(A)} \\ &= - \sum_{B \in \beta} \sum_{A \in \alpha} \mu(A) \frac{\mu(A \cap B)}{\mu(A)} \log \frac{\mu(A \cap B)}{\mu(A)} \\ &= \sum_{B \in \beta} \sum_{A \in \alpha} \mu(A) f\left(\frac{\mu(A \cap B)}{\mu(A)}\right) \\ &\leq \sum_{B \in \beta} f\left(\sum_{A \in \alpha} \mu(A) \frac{\mu(A \cap B)}{\mu(A)}\right) \\ &= \sum_{B \in \beta} f(\mu(B)) = H(\beta). \end{aligned}$$

□

Exercise 5.2.3 Prove the rest of the properties of Proposition 5.2.1

Now consider the partition $\bigvee_{i=0}^{n-1} T^{-i}\alpha$, whose atoms are of the form $A_{i_0} \cap T^{-1}A_{i_1} \cap \dots \cap T^{-(n-1)}A_{i_{n-1}}$, consisting of all points $x \in X$ with the property that $x \in A_{i_0}, Tx \in A_{i_1}, \dots, T^{n-1}x \in A_{i_{n-1}}$.

Exercise 5.2.4 Show that if α is a finite partition of (X, \mathcal{F}, μ, T) , then

$$H\left(\bigvee_{i=0}^{n-1} T^{-i}\alpha\right) = H(\alpha) + \sum_{j=1}^{n-1} H(\alpha | \bigvee_{i=1}^j T^{-i}\alpha).$$

To define the notion of the entropy of a transformation with respect to a partition, we need the following two propositions.

Proposition 5.2.2 If $\{a_n\}$ is a subadditive sequence of real numbers i.e., $a_{n+p} \leq a_n + a_p$ for all n, p , then

$$\lim_{n \rightarrow \infty} \frac{a_n}{n}$$

exists.

Proof. Fix any $m > 0$. For any $n \geq 0$ one has $n = km + i$ for some i between $0 \leq i \leq m - 1$. By subadditivity it follows that

$$\frac{a_n}{n} = \frac{a_{km+i}}{km+i} \leq \frac{a_{km}}{km} + \frac{a_i}{km} \leq k \frac{a_m}{km} + \frac{a_i}{km}.$$

Note that if $n \rightarrow \infty, k \rightarrow \infty$ and so $\limsup_{n \rightarrow \infty} a_n/n \leq a_m/m$. Since m is arbitrary one has

$$\limsup_{n \rightarrow \infty} \frac{a_n}{n} \leq \inf \frac{a_m}{m} \leq \liminf_{n \rightarrow \infty} \frac{a_n}{n}.$$

Therefore $\lim_{n \rightarrow \infty} a_n/n$ exists, and equals $\inf a_n/n$. \square

Proposition 5.2.3 Let α be a finite partitions of (X, \mathcal{B}, μ, T) , where T is a measure preserving transformation. Then, $\lim_{n \rightarrow \infty} \frac{1}{n} H(\bigvee_{i=0}^{n-1} T^{-i}\alpha)$ exists.

Proof. Let $a_n = H(\bigvee_{i=0}^{n-1} T^{-i}\alpha) \geq 0$. Then, by Proposition 5.2.1, we have

$$\begin{aligned} a_{n+p} &= H\left(\bigvee_{i=0}^{n+p-1} T^{-i}\alpha\right) \\ &\leq H\left(\bigvee_{i=0}^{n-1} T^{-i}\alpha\right) + H\left(\bigvee_{i=n}^{n+p-1} T^{-i}\alpha\right) \\ &= a_n + H\left(\bigvee_{i=0}^{p-1} T^{-i}\alpha\right) \\ &= a_n + a_p. \end{aligned}$$

Hence, by Proposition 5.2.2,

$$\lim_{n \rightarrow \infty} \frac{a_n}{n} = \lim_{n \rightarrow \infty} \frac{1}{n} H\left(\bigvee_{i=0}^{n-1} T^{-i}\alpha\right)$$

exists. □

We are now in position to give the definition of the entropy of the transformation T .

Definition 5.2.1 *The entropy of the measure preserving transformation T with respect to the partition α is given by*

$$h(\alpha, T) = h_\mu(\alpha, T) := \lim_{n \rightarrow \infty} \frac{1}{n} H\left(\bigvee_{i=0}^{n-1} T^{-i}\alpha\right),$$

where

$$H\left(\bigvee_{i=0}^{n-1} T^{-i}\alpha\right) = - \sum_{D \in \bigvee_{i=0}^{n-1} T^{-i}\alpha} \mu(D) \log(\mu(D)).$$

Finally, the entropy of the transformation T is given by

$$h(T) = h_\mu(T) := \sup_{\alpha} h(\alpha, T).$$

The following theorem gives an equivalent definition of entropy.

Theorem 5.2.1 *The entropy of the measure preserving transformation T with respect to the partition α is also given by*

$$h(\alpha, T) = \lim_{n \rightarrow \infty} H(\alpha | \bigvee_{i=1}^{n-1} T^{-i}\alpha).$$

Proof. Notice that the sequence $\{H(\alpha | \bigvee_{i=1}^n T^{-i}\alpha)\}_{n \geq 1}$ is bounded from below, and is non-increasing, hence $\lim_{n \rightarrow \infty} H(\alpha | \bigvee_{i=1}^n T^{-i}\alpha)$ exists. Furthermore,

$$\lim_{n \rightarrow \infty} H(\alpha | \bigvee_{i=1}^n T^{-i}\alpha) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=1}^n H(\alpha | \bigvee_{i=1}^j T^{-i}\alpha).$$

From Exercise 5.2.4, we have

$$H(\bigvee_{i=0}^{n-1} T^{-i}\alpha) = H(\alpha) + \sum_{j=1}^{n-1} H(\alpha | \bigvee_{i=1}^j T^{-i}\alpha).$$

Now, dividing by n , and taking the limit as $n \rightarrow \infty$, one gets the desired result \square

Theorem 5.2.2 *Entropy is an isomorphism invariant.*

Proof. Let (X, \mathcal{B}, μ, T) and (Y, \mathcal{C}, ν, S) be two isomorphic measure preserving systems (see Definition 1.2.3, for a definition), with $\psi : X \rightarrow Y$ the corresponding isomorphism. We need to show that $h_\mu(T) = h_\nu(S)$.

Let $\beta = \{B_1, \dots, B_n\}$ be any partition of Y , then

$$\psi^{-1}\beta = \{\psi^{-1}B_1, \dots, \psi^{-1}B_n\}$$

is a partition of X . Set $A_i = \psi^{-1}B_i$, for $1 \leq i \leq n$. Since $\psi : X \rightarrow Y$ is an isomorphism, we have that $\nu = \mu\psi^{-1}$ and $\psi T = S\psi$, so that for any $n \geq 0$ and $B_{i_0}, \dots, B_{i_{n-1}} \in \beta$

$$\begin{aligned} & \nu(B_{i_0} \cap S^{-1}B_{i_1} \cap \dots \cap S^{-(n-1)}B_{i_{n-1}}) \\ &= \mu(\psi^{-1}B_{i_0} \cap \psi^{-1}S^{-1}B_{i_1} \cap \dots \cap \psi^{-1}S^{-(n-1)}B_{i_{n-1}}) \\ &= \mu(\psi^{-1}B_{i_0} \cap T^{-1}\psi^{-1}B_{i_1} \cap \dots \cap T^{-(n-1)}\psi^{-1}B_{i_{n-1}}) \\ &= \mu(A_{i_0} \cap T^{-1}A_{i_1} \cap \dots \cap T^{-(n-1)}A_{i_{n-1}}). \end{aligned}$$

Setting

$$A(n) = A_{i_0} \cap \cdots \cap T^{-(n-1)}A_{i_{n-1}} \quad \text{and} \quad B(n) = B_{i_0} \cap \cdots \cap S^{-(n-1)}B_{i_{n-1}},$$

we thus find that

$$\begin{aligned} h_\nu(S) &= \sup_{\beta} h_\nu(\beta, S) = \sup_{\beta} \lim_{n \rightarrow \infty} \frac{1}{n} H_\nu\left(\bigvee_{i=0}^{n-1} S^{-i}\beta\right) \\ &= \sup_{\beta} \lim_{n \rightarrow \infty} -\frac{1}{n} \sum_{B(n) \in \mathcal{V}_{i=0}^{n-1} S^{-i}\beta} \nu(B(n)) \log \nu(B(n)) \\ &= \sup_{\psi^{-1}\beta} \lim_{n \rightarrow \infty} -\frac{1}{n} \sum_{A(n) \in \mathcal{V}_{i=0}^{n-1} T^{-i}\psi^{-1}\beta} \mu(A(n)) \log \mu(A(n)) \\ &= \sup_{\psi^{-1}\beta} h_\mu(\psi^{-1}\beta, T) \\ &\leq \sup_{\alpha} h_\mu(\alpha, T) = h_\mu(T), \end{aligned}$$

where in the last inequality the supremum is taken over all possible finite partitions α of X . Thus $h_\nu(S) \leq h_\mu(T)$. The proof of $h_\mu(T) \leq h_\nu(S)$ is done similarly. Therefore $h_\nu(S) = h_\mu(T)$, and the proof is complete. \square

5.3 Calculation of Entropy and Examples

Calculating the entropy of a transformation directly from the definition does not seem feasible, for one needs to take the supremum over **all** finite partitions, which is practically impossible. However, the entropy of a partition is relatively easier to calculate if one has full information about the partition under consideration. So the question is whether it is possible to find a partition α of X where $h(\alpha, T) = h(T)$. Naturally, such a partition contains all the information ‘transmitted’ by T . To answer this question we need some notations and definitions.

For $\alpha = \{A_1, \dots, A_N\}$ and all $m, n \geq 0$, let

$$\sigma\left(\bigvee_{i=n}^m T^{-i}\alpha\right) \quad \text{and} \quad \sigma\left(\bigvee_{i=-m}^{-n} T^{-i}\alpha\right)$$

be the smallest σ -algebras containing the partitions $\bigvee_{i=n}^m T^{-i}\alpha$ and $\bigvee_{i=-m}^{-n} T^{-i}\alpha$ respectively. Furthermore, let $\sigma\left(\bigvee_{i=-\infty}^{-\infty} T^{-i}\alpha\right)$ be the smallest σ -algebra containing all the partitions $\bigvee_{i=n}^m T^{-i}\alpha$ and $\bigvee_{i=-m}^{-n} T^{-i}\alpha$ for all n and m . We

call a partition α a *generator* with respect to T if $\sigma(\bigvee_{i=-\infty}^{\infty} T^{-i}\alpha) = \mathcal{F}$, where \mathcal{F} is the σ -algebra on X . If T is non-invertible, then α is said to be a generator if $\sigma(\bigvee_{i=0}^{\infty} T^{-i}\alpha) = \mathcal{F}$. Naturally, this equality is modulo sets of measure zero. One has also the following characterization of generators, saying basically, that each measurable set in X can be approximated by a finite disjoint union of cylinder sets. See also [W] for more details and proofs.

Proposition 5.3.1 *The partition α is a generator of \mathcal{F} if for each $A \in \mathcal{F}$ and for each $\varepsilon > 0$ there exists a finite disjoint union C of elements of $\{\alpha_n^m\}$, such that $\mu(A\Delta C) < \varepsilon$.*

We now state (without proofs) two famous theorems known as *Kolmogorov-Sinai's Theorem* and *Krieger's Generator Theorem*. For the proofs, we refer the interested reader to the book of Karl Petersen ([P]) or Peter Walters ([W]).

Theorem 5.3.1 (Kolmogorov and Sinai, 1958) *If α is a generator with respect to T and $H(\alpha) < \infty$, then $h(T) = h(\alpha, T)$.*

Theorem 5.3.2 (Krieger, 1970) *If T is an ergodic measure preserving transformation with $h(T) < \infty$, then T has a finite generator.*

We will use these two theorems to calculate the entropy of a Bernoulli shift.

Example (Entropy of a Bernoulli shift)—Let T be the left shift on $X = \{1, 2, \dots, n\}^{\mathbb{Z}}$ endowed with the σ -algebra \mathcal{F} generated by the cylinder sets, and product measure μ giving symbol i probability p_i , where $p_1 + p_2 + \dots + p_n = 1$. Our aim is to calculate $h(T)$. To this end we need to find a partition α which generates the σ -algebra \mathcal{F} under the action of T . The natural choice of α is what is known as the *time-zero partition* $\alpha = \{A_1, \dots, A_n\}$, where

$$A_i := \{x \in X : x_0 = i\}, \quad i = 1, \dots, n.$$

Notice that for all $m \in \mathbb{Z}$,

$$T^{-m}A_i = \{x \in X : x_m = i\},$$

and

$$A_{i_0} \cap T^{-1}A_{i_1} \cap \dots \cap T^{-m}A_{i_m} = \{x \in X : x_0 = i_0, \dots, x_m = i_m\}.$$

In other words, $\bigvee_{i=0}^m T^{-i}\alpha$ is precisely the collection of cylinders of length m (i.e., the collection of all m -blocks), and these by definition generate \mathcal{F} . Hence, α is a generating partition, so that

$$h(T) = h(\alpha, T) = \lim_{m \rightarrow \infty} \frac{1}{m} H \left(\bigvee_{i=0}^{m-1} T^{-i}\alpha \right).$$

First notice that – since μ is product measure here – the partitions

$$\alpha, T^{-1}\alpha, \dots, T^{-(m-1)}\alpha$$

are all independent since each specifies a different coordinate, and so

$$\begin{aligned} H(\alpha \vee T^{-1}\alpha \vee \dots \vee T^{-(m-1)}\alpha) &= H(\alpha) + H(T^{-1}\alpha) + \dots + H(T^{-(m-1)}\alpha) \\ &= mH(\alpha) = -m \sum_{i=1}^n p_i \log p_i. \end{aligned}$$

Thus,

$$h(T) = \lim_{m \rightarrow \infty} \frac{1}{m} (-m) \sum_{i=1}^n p_i \log p_i = - \sum_{i=1}^n p_i \log p_i.$$

Exercise 5.3.1 Let T be the left shift on $X = \{1, 2, \dots, n\}^{\mathbb{Z}}$ endowed with the σ -algebra \mathcal{F} generated by the cylinder sets, and the Markov measure μ given by the stochastic matrix $P = (p_{ij})$, and the probability vector $\pi = (\pi_1, \dots, \pi_n)$ with $\pi P = \pi$. Show that

$$h(T) = - \sum_{j=1}^n \sum_{i=1}^n \pi_i p_{ij} \log p_{ij}$$

Exercise 5.3.2 Let T be a measure preserving transformation on the probability space (X, \mathcal{F}, μ) .

- (a) Suppose α is a finite partition of X . Show that $h_\mu(\alpha, T) = h_\mu(\bigvee_{i=1}^n T^{-i}\alpha, T)$, for any $n \geq 1$.

- (b) Let α and β be finite partitions. Show that $h_\mu(\beta, T) \leq h_\mu(\alpha, T) + H_\mu(\beta|\alpha)$.
- (c) Suppose α is a finite generator, i.e. $\sigma(\bigvee_{i=0}^{\infty} T^{-i}\alpha) = \mathcal{F}$. Using parts (a) and (b), show that for any finite partition β of X one has $h_\mu(\beta, T) \leq h_\mu(\alpha, T)$. Conclude that $h_\mu(\alpha, T) = h_\mu(T)$.

Exercise 5.3.3 Let (X, \mathcal{F}, μ) be a probability space and $T : X \rightarrow X$ a measure preserving transformation. Let $k > 0$.

- (a) Show that for any finite partition α of X one has $h_\mu(\bigvee_{i=0}^{k-1} T^{-i}\alpha, T^k) = kh_\mu(\alpha, T)$.
- (b) Prove that $kh_\mu(T) \leq h_\mu(T^k)$.
- (c) Prove that $h_\mu(\alpha, T^k) \leq kh_\mu(\alpha, T)$.
- (d) Prove that $h_\mu(T^k) = kh_\mu(T)$.

Throughout this course (label mentioned), the spaces we are working with are assumed to be *standard Lebesgue spaces*. This means that one has a measurable isomorphism (up to sets of measure zero) between our probability space and the unit interval with possibly finite or countable number of atoms. For such spaces it is possible to find an increasing sequence of finite partitions $\alpha_1 \leq \alpha_2 \leq \dots$ such that $\sigma(\bigvee_n \alpha_n) = \mathcal{F}$. This allows us to calculate entropy in the following way.

Lemma 5.3.1 *If $\alpha_1 \leq \alpha_2 \leq \dots$ is an increasing sequence of finite partitions on (X, \mathcal{F}, μ, T) such $\sigma(\alpha_n) \nearrow \mathcal{F}$, then $h_\mu(T) = \lim_{n \rightarrow \infty} h_\mu(\alpha_n, T)$.*

Proof It is enough to show that for any finite partition β , one has $h_\mu(\beta, T) \leq \lim_{n \rightarrow \infty} h_\mu(\alpha_n, T)$. From Exercise 5.3.2(b),

$$h_\mu(\beta, T) \leq h_\mu(\alpha_n, T) + H(\beta|\alpha_n).$$

Since $\sigma(\alpha_n) \nearrow \mathcal{F}$, then by the Martingale Convergence Theorem (see Theorem 5.4.1) together with the Dominated Convergence Theorem one has

$$\lim_{n \rightarrow \infty} H(\beta|\alpha_n) = \lim_{n \rightarrow \infty} H(\beta|\sigma(\alpha_n)) = H(\beta|\mathcal{F}) = 0.$$

Thus, $h_\mu(\beta, T) \leq \lim_{n \rightarrow \infty} h_\mu(\alpha_n, T)$, and hence $h_\mu(T) = \lim_{n \rightarrow \infty} h_\mu(\alpha_n, T)$. \square

Exercise 5.3.4 Prove that $h_{\mu_1 \times \mu_2}(T_1 \times T_2) = h_{\mu_1}(T_1) + h_{\mu_2}(T_2)$.

5.4 The Shannon-McMillan-Breiman Theorem

In the previous sections we have considered only finite partitions on X , however all the definitions and results hold if we were to consider countable partitions of finite entropy. Before we state and prove the Shannon-McMillan-Breiman Theorem, we need to introduce the information function associated with a partition.

Let (X, \mathcal{F}, μ) be a probability space, and $\alpha = \{A_1, A_2, \dots\}$ be a finite or a countable partition of X into measurable sets. For each $x \in X$, let $\alpha(x)$ be the element of α to which x belongs. Then, the *information function* associated to α is defined to be

$$I_\alpha(x) = -\log \mu(\alpha(x)) = -\sum_{A \in \alpha} 1_A(x) \log \mu(A).$$

For two finite or countable partitions α and β of X , we define the *conditional information function* of α given β by

$$I_{\alpha|\beta}(x) = -\sum_{B \in \beta} \sum_{A \in \alpha} 1_{(A \cap B)}(x) \log \left(\frac{\mu(A \cap B)}{\mu(B)} \right).$$

We claim that

$$I_{\alpha|\beta}(x) = -\log E_\mu(1_{\alpha(x)} | \sigma(\beta)) = -\sum_{A \in \alpha} 1_A(x) \log E(1_A | \sigma(\beta)), \quad (5.3)$$

where $\sigma(\beta)$ is the σ -algebra generated by the finite or countable partition β , (see the remark following the proof of Theorem (2.1.1)). This follows from the fact (which is easy to prove using the definition of conditional expectations) that if β is finite or countable, then for any $f \in L^1(\mu)$, one has

$$E_\mu(f | \sigma(\beta)) = \sum_{B \in \beta} 1_B \frac{1}{\mu(B)} \int_B f d\mu.$$

Clearly, $H(\alpha|\beta) = \int_X I_{\alpha|\beta}(x) d\mu(x)$.

Exercise 5.4.1 Let α and β be finite or countable partitions of X . Show that $I_{\alpha \vee \beta} = I_{\alpha} + I_{\beta|\alpha}$.

Now suppose $T : X \rightarrow X$ is a measure preserving transformation on (X, \mathcal{F}, μ) , and let $\alpha = \{A_1, A_2, \dots\}$ be any countable partition. Then $T^{-1} = \{T^{-1}A_1, T^{-1}A_2, \dots\}$ is also a countable partition. Since T is measure preserving one has,

$$I_{T^{-1}\alpha}(x) = - \sum_{A_i \in \alpha} 1_{T^{-1}A_i}(x) \log \mu(T^{-1}A_i) = - \sum_{A_i \in \alpha} 1_{A_i}(Tx) \log \mu(A_i) = I_{\alpha}(Tx).$$

Furthermore,

$$\lim_{n \rightarrow \infty} \frac{1}{n+1} H\left(\bigvee_{i=0}^n T^{-i}\alpha\right) = \lim_{n \rightarrow \infty} \frac{1}{n+1} \int_X I_{\bigvee_{i=0}^n T^{-i}\alpha}(x) d\mu(x) = h(\alpha, T).$$

The Shannon-McMillan-Breiman theorem says if T is ergodic and if α has finite entropy, then in fact the integrand $\frac{1}{n+1} I_{\bigvee_{i=0}^n T^{-i}\alpha}(x)$ converges a.e. to $h(\alpha, T)$. Notice that the integrand can be written as

$$\frac{1}{n+1} I_{\bigvee_{i=0}^n T^{-i}\alpha}(x) = -\frac{1}{n+1} \log \mu\left(\bigvee_{i=0}^n T^{-i}\alpha(x)\right),$$

where $(\bigvee_{i=0}^n T^{-i}\alpha)(x)$ is the element of $\bigvee_{i=0}^n T^{-i}\alpha$ containing x (often referred to as the α -cylinder of order n containing x). Before we proceed we need the following proposition.

Proposition 5.4.1 Let $\alpha = \{A_1, A_2, \dots\}$ be a countable partition with finite entropy. For each $n = 1, 2, 3, \dots$, let $f_n(x) = I_{\alpha|\bigvee_{i=1}^n T^{-i}\alpha}(x)$, and let $f^* = \sup_{n \geq 1} f_n$. Then, for each $t \geq 0$ and for each $A \in \alpha$,

$$\mu(\{x \in A : f^*(x) > t\}) \leq 2^{-t}.$$

Furthermore, $f^* \in L^1(X, \mathcal{F}, \mu)$.

Proof. Let $t \geq 0$ and $A \in \alpha$. For $n \geq 1$, let

$$f_n^A(x) = -\log E_{\mu} \left(1_A \middle| \bigvee_{i=1}^n T^{-i}\alpha \right) (x),$$

and

$$B_n = \{x \in X : f_1^A(x) \leq t, \dots, f_{n-1}^A(x) \leq t, f_n^A(x) > t\}.$$

Notice that for $x \in A$ one has $f_n(x) = f_n^A(x)$, and for $x \in B_n$ one has $E_\mu(1_A | \bigvee_{i=1}^n T^{-i}\alpha)(x) < 2^{-t}$. Since $B_n \in \sigma(\bigvee_{i=1}^n T^{-i}\alpha)$, then

$$\begin{aligned} \mu(B_n \cap A) &= \int_{B_n} 1_A(x) \, d\mu(x) \\ &= \int_{B_n} E_\mu \left(1_A | \bigvee_{i=1}^n T^{-i}\alpha \right) (x) \, d\mu(x) \\ &\leq \int_{B_n} 2^{-t} \, d\mu(x) = 2^{-t} \mu(B_n). \end{aligned}$$

Thus,

$$\begin{aligned} \mu(\{x \in A : f^*(x) > t\}) &= \mu(\{x \in A : f_n(x) > t, \text{ for some } n\}) \\ &= \mu(\{x \in A : f_n^A(x) > t, \text{ for some } n\}) \\ &= \mu(\cup_{n=1}^{\infty} A \cap B_n) \\ &= \sum_{n=1}^{\infty} \mu(A \cap B_n) \\ &\leq 2^{-t} \sum_{n=1}^{\infty} \mu(B_n) \leq 2^{-t}. \end{aligned}$$

We now show that $f^* \in L^1(X, \mathcal{F}, \mu)$. First notice that

$$\mu(\{x \in A : f^*(x) > t\}) \leq \mu(A),$$

hence,

$$\mu(\{x \in A : f^*(x) > t\}) \leq \min(\mu(A), 2^{-t}).$$

Using Fubini's Theorem, and the fact that $f^* \geq 0$ one has

$$\begin{aligned}
\int_X f^*(x) d\mu(x) &= \int_0^\infty \mu(\{x \in X : f^*(x) > t\}) dt \\
&= \int_0^\infty \sum_{A \in \alpha} \mu(\{x \in A : f^*(x) > t\}) dt \\
&= \sum_{A \in \alpha} \int_0^\infty \mu(\{x \in A : f^*(x) > t\}) dt \\
&\leq \sum_{A \in \alpha} \int_0^\infty \min(\mu(A), 2^{-t}) dt \\
&= \sum_{A \in \alpha} \int_0^{-\log \mu(A)} \mu(A) dt + \sum_{A \in \alpha} \int_{-\log \mu(A)}^\infty 2^{-t} dt \\
&= -\sum_{A \in \alpha} \mu(A) \log \mu(A) + \sum_{A \in \alpha} \frac{\mu(A)}{\log_e 2} \\
&= H_\mu(\alpha) + \frac{1}{\log_e 2} < \infty.
\end{aligned}$$

□

So far we have defined the notion of the conditional entropy $I_{\alpha|\beta}$ when α and β are countable partitions. We can generalize the definition to the case α is a countable partition and \mathcal{G} is a σ -algebra as follows (see equation (5.3)),

$$I_{\alpha|\mathcal{G}}(x) = -\log E_\mu(1_{\alpha(x)}|\mathcal{G}).$$

If we denote by $\bigvee_{i=1}^\infty T^{-i}\alpha = \sigma(\cup_n \bigvee_{i=1}^n T^{-i}\alpha)$, then

$$I_{\alpha|\bigvee_{i=1}^\infty T^{-i}\alpha}(x) = \lim_{n \rightarrow \infty} I_{\alpha|\bigvee_{i=1}^n T^{-i}\alpha}(x). \quad (5.4)$$

Exercise 5.4.2 Give a proof of equation (5.4) using the following important theorem, known as the Martingale Convergence Theorem (and is stated to our setting)

Theorem 5.4.1 (Martingale Convergence Theorem) *Let $\mathcal{C}_1 \subseteq \mathcal{C}_2 \subseteq \dots$ be a sequence of increasing σ -algebras, and let $\mathcal{C} = \sigma(\cup_n \mathcal{C}_n)$. If $f \in L^1(\mu)$, then*

$$E_\mu(f|\mathcal{C}) = \lim_{n \rightarrow \infty} E_\mu(f|\mathcal{C}_n)$$

μ a.e., and in $L^1(\mu)$.

Exercise 5.4.3 Show that if T is measure preserving on the probability space (X, \mathcal{F}, μ) and $f \in L^1(\mu)$, then

$$\lim_{n \rightarrow \infty} \frac{f(T^n x)}{n} = 0, \quad \mu \text{ a.e.}$$

Theorem 5.4.2 (The Shannon-McMillan-Breiman Theorem) Suppose T is an ergodic measure preserving transformation on a probability space (X, \mathcal{F}, μ) , and let α be a countable partition with $H(\alpha) < \infty$. Then,

$$\lim_{n \rightarrow \infty} \frac{1}{n+1} I_{\bigvee_{i=0}^n T^{-i}\alpha}(x) = h(\alpha, T) \text{ a.e.}$$

Proof. For each $n = 1, 2, 3, \dots$, let $f_n(x) = I_{\alpha|_{\bigvee_{i=1}^n T^{-i}\alpha}}(x)$. Then,

$$\begin{aligned} I_{\bigvee_{i=0}^n T^{-i}\alpha}(x) &= I_{\bigvee_{i=1}^n T^{-i}\alpha}(x) + I_{\alpha|_{\bigvee_{i=1}^n T^{-i}\alpha}}(x) \\ &= I_{\bigvee_{i=0}^{n-1} T^{-i}\alpha}(Tx) + f_n(x) \\ &= I_{\bigvee_{i=1}^{n-1} T^{-i}\alpha}(T^2x) + I_{\alpha|_{\bigvee_{i=1}^{n-1} T^{-i}\alpha}}(T^2x) + f_n(x) \\ &= I_{\bigvee_{i=0}^{n-2} T^{-i}\alpha}(T^2x) + f_{n-1}(Tx) + f_n(x) \\ &\vdots \\ &= I_{\alpha}(T^n x) + f_1(T^{n-1}x) + \dots + f_{n-1}(Tx) + f_n(x). \end{aligned}$$

Let $f(x) = I_{\alpha|_{\bigvee_{i=1}^{\infty} T^{-i}\alpha}}(x) = \lim_{n \rightarrow \infty} f_n(x)$. Notice that $f \in L^1(X, \mathcal{F}, \mu)$ since $\int_X f(x) d\mu(x) = h(\alpha, T)$. Now letting $f_0 = I_{\alpha}$, we have

$$\begin{aligned} \frac{1}{n+1} I_{\bigvee_{i=0}^n T^{-i}\alpha}(x) &= \frac{1}{n+1} \sum_{k=0}^n f_{n-k}(T^k x) \\ &= \frac{1}{n+1} \sum_{k=0}^n f(T^k x) + \frac{1}{n+1} \sum_{k=0}^n (f_{n-k} - f)(T^k x). \end{aligned}$$

By the ergodic theorem,

$$\lim_{n \rightarrow \infty} \sum_{k=0}^n f(T^k x) = \int_X f(x) d\mu(x) = h(\alpha, T) \text{ a.e.}$$

We now study the sequence $\left\{ \frac{1}{n+1} \sum_{k=0}^n (f_{n-k} - f)(T^k x) \right\}$. Let

$$F_N = \sup_{k \geq N} |f_k - f|, \quad \text{and} \quad f^* = \sup_{n \geq 1} f_n.$$

Notice that $0 \leq F_N \leq f^* + f$, hence $F_N \in L^1(X, \mathcal{F}, \mu)$ and $\lim_{N \rightarrow \infty} F_N(x) = 0$ a.e. By the Lebesgue Dominated Convergence Theorem, one has $\lim_{N \rightarrow \infty} \int F_N d\mu$ is equal to 0. Also for any k , $|f_{n-k} - f| \leq f^* + f$, so that $|f_{n-k} - f| \in L^1(X, \mathcal{F}, \mu)$ and $\lim_{n \rightarrow \infty} |f_{n-k} - f| = 0$ a.e.

For any $N \geq 1$, and for all $n \geq N$ one has

$$\begin{aligned} \frac{1}{n+1} \sum_{k=0}^n |f_{n-k} - f|(T^k x) &= \frac{1}{n+1} \sum_{k=0}^{n-N} |f_{n-k} - f|(T^k x) \\ &\quad + \frac{1}{n+1} \sum_{k=n-N+1}^n |f_{n-k} - f|(T^k x) \\ &\leq \frac{1}{n+1} \sum_{k=0}^{n-N} F_N(T^k x) \\ &\quad + \frac{1}{n+1} \sum_{k=0}^{N-1} |f_k - f|(T^{n-k} x). \end{aligned}$$

If we take the limit as $n \rightarrow \infty$, then by Exercise 5.4.3, the second term tends to 0 a.e. and by the Ergodic Theorem the first term tends to $\int F_N d\mu$. Now taking the limit as $N \rightarrow \infty$, one sees that

$$\frac{1}{n+1} \sum_{k=0}^n |f_{n-k} - f|(T^k x) = 0 \quad \text{a.e.}$$

Hence,

$$\lim_{n \rightarrow \infty} \frac{1}{n+1} I_{\bigvee_{i=0}^n T^{-i} \alpha}(x) = h(\alpha, T) \quad \text{a.e.}$$

□

The above theorem can be interpreted as providing an estimate of the size of the atoms of $\bigvee_{i=0}^n T^{-i} \alpha$. For n sufficiently large, a typical element $A \in \bigvee_{i=0}^n T^{-i} \alpha$ satisfies

$$-\frac{1}{n+1} \log \mu(A) \approx h(\alpha, T)$$

or

$$\mu(A_n) \approx 2^{-(n+1)h(\alpha, T)}.$$

Furthermore, if α is a generating partition (i.e., $\bigvee_{i=0}^{\infty} T^{-i}\alpha = \mathcal{F}$), then in the conclusion of Shannon-McMillan-Breiman Theorem one can replace $h(\alpha, T)$ by $h(T)$.

Exercise 5.4.4 Let (X, \mathcal{F}, μ, T) be a measure preserving and ergodic system. Suppose α is a partition of X into measurable sets. For $x \in X$, let $\alpha_n(x)$ be the element of the partition $\bigvee_{i=0}^{n-1} T^{-i}\alpha$ that contains x . Suppose that there exist constants $0 < C_1 < C_2$ such that $C_1\lambda(A) < \mu(A) < C_2\lambda(A)$ for all $A \in \mathcal{F}$. Show that the conclusion of the Shannon-McMillan-Breiman Theorem holds if we replace μ by λ , i.e.

$$\lim_{n \rightarrow \infty} -\frac{\log \lambda(\alpha_n(x))}{n} = h_\mu(\alpha, T) \text{ a.e. with respect to } \lambda.$$

Exercise 5.4.5 let $T : [0, 1) \rightarrow [0, 1)$ be given by $Tx = \beta x \pmod{1}$, where $\beta = \frac{1 + \sqrt{5}}{2}$. Use Shannon McMillan Breiman Theorem and Exercise 5.4.4 to calculate the entropy $h_\mu(T)$ of T with respect to the invariant measure μ given by

$$\mu(A) = \int_A g(x) d\lambda(x),$$

with

$$g(x) = \begin{cases} \frac{5 + 3\sqrt{5}}{10}, & 0 \leq x < \frac{\sqrt{5} - 1}{2} \\ \frac{5 + \sqrt{5}}{10}, & 0 \leq \frac{\sqrt{5} - 1}{2} \leq x < 1. \end{cases}$$

Exercise 5.4.6 Use Shannon-McMillan-Breiman Theorem, Sinai Kolmogorov Theorem and Exercise 5.4.4 to show that if T is the continued fraction map and μ is Gauss measure, then $h_\mu(T) = \frac{\pi^2}{6 \log 2}$.

5.5 Lochs' Theorem

In 1964, G. Lochs compared the decimal and the continued fraction expansions. Let $x \in [0, 1)$ be an irrational number, and suppose $x = .d_1d_2\dots$ is the decimal expansion of x (which is generated by iterating the map $Sx = 10x \pmod{1}$). Suppose further that

$$x = \frac{1}{a_1 + \frac{1}{a_2 + \frac{1}{a_3 + \frac{1}{\ddots}}}} = [0; a_1, a_2, \dots] \quad (5.5)$$

is its regular continued fraction (RCF) expansion (generated by the map $Tx = \frac{1}{x} - \lfloor \frac{1}{x} \rfloor$). Let $y = .d_1d_2\dots d_n$ be the rational number determined by the first n decimal digits of x , and let $z = y + 10^{-n}$. Then, $[y, z)$ is the decimal cylinder of order n containing x , which we also denote by $B_n(x)$. Now let

$$y = \frac{1}{b_1 + \frac{1}{b_2 + \frac{1}{\ddots + \frac{1}{b_l}}}}$$

and

$$z = \frac{1}{c_1 + \frac{1}{c_2 + \frac{1}{\ddots + \frac{1}{c_k}}}}$$

be the continued fraction expansion of y and z . Let

$$m(n, x) = \max \{i \leq \min(l, k) : \text{for all } j \leq i, b_j = c_j\}. \quad (5.6)$$

In other words, if $B_n(x)$ denotes the decimal cylinder consisting of all points y in $[0, 1)$ such that the first n decimal digits of y agree with those of x , and if $C_j(x)$ denotes the continued fraction cylinder of order j containing x , i.e., $C_j(x)$ is the set of all points in $[0, 1)$ such that the first j digits in their continued fraction expansion is the same as that of x , then $m(n, x)$ is the largest integer such that $B_n(x) \subset C_{m(n,x)}(x)$. Lochs proved the following theorem:

Theorem 5.5.1 *Let λ denote Lebesgue measure on $[0, 1)$. Then for a.e. $x \in [0, 1)$*

$$\lim_{n \rightarrow \infty} \frac{m(n, x)}{n} = \frac{6 \log 2 \log 10}{\pi^2}.$$

In this section, we will prove a generalization of Lochs' theorem that allows one to compare any two known expansions of numbers. We show that Lochs' theorem is true for any two sequences of interval partitions on $[0, 1)$ satisfying the conclusion of Shannon-McMillan-Breiman theorem. The content of this section as well as the proofs can be found in [DF]. We begin with few definitions that will be used in the arguments to follow.

Definition 5.5.1 *By an interval partition we mean a finite or countable partition of $[0, 1)$ into subintervals. If P is an interval partition and $x \in [0, 1)$, we let $P(x)$ denote the interval of P containing x .*

Let $\mathcal{P} = \{P_n\}_{n=1}^{\infty}$ be a sequence of interval partitions. Let λ denote Lebesgue probability measure on $[0, 1)$.

Definition 5.5.2 *Let $c \geq 0$. We say that \mathcal{P} has entropy c a.e. with respect to λ if*

$$-\frac{\log \lambda(P_n(x))}{n} \rightarrow c \quad \text{a.e.}$$

Note that we do not assume that each P_n is refined by P_{n+1} .

Suppose that $\mathcal{P} = \{P_n\}_{n=1}^{\infty}$ and $\mathcal{Q} = \{Q_n\}_{n=1}^{\infty}$ are sequences of interval partitions. For each $n \in \mathbb{N}$ and $x \in [0, 1)$, define

$$m_{\mathcal{P}, \mathcal{Q}}(n, x) = \sup \{m \mid P_n(x) \subset Q_m(x)\}.$$

Theorem 5.5.2 *Let $\mathcal{P} = \{P_n\}_{n=1}^{\infty}$ and $\mathcal{Q} = \{Q_n\}_{n=1}^{\infty}$ be sequences of interval partitions and λ Lebesgue probability measure on $[0, 1)$. Suppose that for some constants $c > 0$ and $d > 0$, \mathcal{P} has entropy c a.e with respect to λ and \mathcal{Q} has entropy d a.e. with respect to λ . Then*

$$\frac{m_{\mathcal{P}, \mathcal{Q}}(n, x)}{n} \rightarrow \frac{c}{d} \quad \text{a.e.}$$

Proof. First we show that

$$\limsup_{n \rightarrow \infty} \frac{m_{\mathcal{P}, \mathcal{Q}}(n, x)}{n} \leq \frac{c}{d} \text{ a.e.}$$

Fix $\varepsilon > 0$. Let $x \in [0, 1)$ be a point at which the convergence conditions of the hypotheses are met. Fix $\eta > 0$ so that $\frac{c + \eta}{c - \frac{\eta}{d}} < 1 + \varepsilon$. Choose N so that for all $n \geq N$

$$\lambda(P_n(x)) > 2^{-n(c+\eta)}$$

and

$$\lambda(Q_n(x)) < 2^{-n(d-\eta)}.$$

Fix n so that $\min\left\{n, \frac{c}{d}n\right\} \geq N$, and let m' denote any integer greater than $(1 + \varepsilon)\frac{c}{d}n$. By the choice of η ,

$$\lambda(P_n(x)) > \lambda(Q_{m'}(x))$$

so that $P_n(x)$ is not contained in $Q_{m'}(x)$. Therefore

$$m_{\mathcal{P}, \mathcal{Q}}(n, x) \leq (1 + \varepsilon)\frac{c}{d}n$$

and so

$$\limsup_{n \rightarrow \infty} \frac{m_{\mathcal{P}, \mathcal{Q}}(n, x)}{n} \leq (1 + \varepsilon)\frac{c}{d} \text{ a.e.}$$

Since $\varepsilon > 0$ was arbitrary, we have the desired result.

Now we show that

$$\liminf_{n \rightarrow \infty} \frac{m_{\mathcal{P}, \mathcal{Q}}(n, x)}{n} \geq \frac{c}{d} \text{ a.e.}$$

Fix $\varepsilon \in (0, 1)$. Choose $\eta > 0$ so that $\zeta =: \varepsilon c - \eta\left(1 + (1 - \varepsilon)\frac{c}{d}\right) > 0$. For each $n \in \mathbb{N}$ let $\bar{m}(n) = \left\lfloor (1 - \varepsilon)\frac{c}{d}n \right\rfloor$. For brevity, for each $n \in \mathbb{N}$ we call an element of P_n (respectively Q_n) (n, η) -good if

$$\lambda(P_n(x)) < 2^{-n(c-\eta)}$$

(respectively

$$\lambda(Q_n(x)) > 2^{-n(d+\eta)}.)$$

For each $n \in \mathbb{N}$, let

$$D_n(\eta) = \left\{ x : \begin{array}{l} P_n(x) \text{ is } (n, \eta) \text{ - good and } Q_{\bar{m}(n)}(x) \text{ is } (\bar{m}(n), \eta) \text{ - good} \\ \text{and } P_n(x) \not\subseteq Q_{\bar{m}(n)}(x) \end{array} \right\}.$$

If $x \in D_n(\eta)$, then $P_n(x)$ contains an endpoint of the $(\bar{m}(n), \eta)$ -good interval $Q_{\bar{m}(n)}(x)$. By the definition of $D_n(\eta)$ and $\bar{m}(n)$,

$$\frac{\lambda(P_n(x))}{\lambda(Q_{\bar{m}(n)}(x))} < 2^{-n\zeta}.$$

Since no more than one atom of P_n can contain a particular endpoint of an atom of $Q_{\bar{m}(n)}$, we see that $\lambda(D_n(\eta)) < 2 \cdot 2^{-n\zeta}$ and so

$$\sum_{n=1}^{\infty} \lambda(D_n(\eta)) < \infty.$$

By Borel-Cantelli, this implies that

$$\lambda\{x \mid x \in D_n(\eta) \text{ i.o.}\} = 0.$$

Since $\bar{m}(n)$ goes to infinity as n does, we have shown that for almost every $x \in [0, 1)$, there exists $N \in \mathbb{N}$, so that for all $n \geq N$, $P_n(x)$ is (n, η) -good and $Q_{\bar{m}(n)}(x)$ is $(\bar{m}(n), \eta)$ -good and $x \notin D_n(\eta)$. In other words, for almost every $x \in [0, 1)$, there exists $N \in \mathbb{N}$, so that for all $n \geq N$, $P_n(x)$ is (n, η) -good and $Q_{\bar{m}(n)}(x)$ is $(\bar{m}(n), \eta)$ -good and $P_n(x) \subset Q_{\bar{m}(n)}(x)$. Thus, for almost every $x \in [0, 1)$, there exists $N \in \mathbb{N}$, so that for all $n \geq N$, $m_{\mathcal{P}, \mathcal{Q}}(n, x) \geq \bar{m}(n)$, so that

$$\frac{m_{\mathcal{P}, \mathcal{Q}}(n, x)}{n} \geq \lfloor (1 - \varepsilon) \frac{c}{d} \rfloor.$$

This proves that

$$\liminf_{n \rightarrow \infty} \frac{m_{\mathcal{P}, \mathcal{Q}}(n, x)}{n} \geq (1 - \varepsilon) \frac{c}{d} \text{ a.e.}$$

Since $\varepsilon > 0$ was arbitrary, we have established the theorem. \square

The above result allows us to compare any two well-known expansions of numbers. Since the *commonly used* expansions are usually performed for points in the unit interval, our underlying space will be $([0, 1), \mathcal{B}, \lambda)$, where \mathcal{B} is the Lebesgue σ -algebra, and λ the Lebesgue measure. The expansions we have in mind share the following two properties.

Definition 5.5.3 A surjective map $T : [0, 1) \rightarrow [0, 1)$ is called a number theoretic fibred map (NTFM) if it satisfies the following conditions:

- (a) there exists a finite or countable partition of intervals $\alpha = \{A_i; i \in D\}$ such that T restricted to each atom of α (cylinder set of order 0) is monotone, continuous and injective. Furthermore, α is a generating partition.
- (b) T is ergodic with respect to Lebesgue measure λ , and there exists a T invariant probability measure μ equivalent to λ with bounded density. (Both $\frac{d\mu}{d\lambda}$ and $\frac{d\lambda}{d\mu}$ are bounded, and $\mu(A) = 0$ if and only if $\lambda(A) = 0$ for all Lebesgue sets A .)

Iterations of T generate expansions of points $x \in [0, 1)$ with digits in D . We refer to the resulting expansion as the T -expansion of x . Almost all known expansions on $[0, 1)$ are generated by a NTFM. Among them are the n -adic expansions ($Tx = nx \pmod{1}$, where n is a positive integer), β expansions ($Tx = \beta x \pmod{1}$, where $\beta > 1$ is a real number), continued fraction expansions ($Tx = \frac{1}{x} - \lfloor \frac{1}{x} \rfloor$), and many others, like [DK].

Let T be an NTFM with corresponding partition α , and T -invariant measure μ equivalent to λ . Let $L, M > 0$ be such that

$$L\lambda(A) \leq \mu(A) < M\lambda(A)$$

for all Lebesgue sets A (property (b)). For $n \geq 1$, let $P_n = \bigvee_{i=0}^{n-1} T^{-i}\alpha$, then by property (a), P_n is an interval partition. If $H_\mu(\alpha) < \infty$, then Shannon-McMillan-Breiman Theorem gives

$$\lim_{n \rightarrow \infty} -\frac{\log \mu(P_n(x))}{n} = h_\mu(T) \quad \text{a.e. with respect to } \mu.$$

Exercise 5.5.1 Prove Theorem 5.5.1 using Theorem 5.5.2. Use the fact that the continued fraction map T is ergodic with respect to Gauss measure μ , given by

$$\mu(B) = \int_B \frac{1}{\log 2} \frac{1}{1+x} dx,$$

and has entropy equal to $h_\mu(T) = \frac{\pi^2}{6 \log 2}$.

Chapter 6

Invariant Measures for Continuous Transformations

6.1 Existence

Suppose X is a compact metric space, and let \mathcal{B} be the Borel σ -algebra i.e., the σ -algebra generated by the open sets. Let $M(X)$ be the collection of all Borel probability measures on X . There is natural embedding of the space X in $M(X)$ via the map $x \rightarrow \delta_x$, where δ_x is the Dirac measure concentrated at x ($\delta_x(A) = 1$ if $x \in A$, and is zero otherwise). Furthermore, $M(X)$ is a convex set, i.e., $p\mu + (1-p)\nu \in M(X)$ whenever $\mu, \nu \in M(X)$ and $0 \leq p \leq 1$. Theorem 6.1.2 below shows that a member of $M(X)$ is determined by how it integrates continuous functions. We denote by $C(X)$ the Banach space of all complex valued continuous functions on X under the supremum norm.

Theorem 6.1.1 *Every member of $M(X)$ is regular, i.e., for all $B \in \mathcal{B}$ and every $\epsilon > 0$ there exist an open set U_ϵ and a closed set C_ϵ such that $C_\epsilon \subseteq B \subseteq U_\epsilon$ such that $\mu(U_\epsilon \setminus C_\epsilon) < \epsilon$.*

Idea of proof. Call a set $B \in \mathcal{B}$ with the above property a *regular set*. Let $\mathcal{R} = \{B \in \mathcal{B} : B \text{ is regular}\}$. Show that \mathcal{R} is a σ -algebra containing all the closed sets. \square

Corollary 6.1.1 *For any $B \in \mathcal{B}$, and any $\mu \in M(X)$,*

$$\mu(B) = \sup_{C \subseteq B: C \text{ closed}} \mu(C) = \inf_{B \subseteq U: U \text{ open}} \mu(U).$$

Theorem 6.1.2 *Let $\mu, m \in M(X)$. If*

$$\int_X f \, d\mu = \int_X f \, dm$$

for all $f \in C(X)$, then $\mu = m$.

Proof. From the above corollary, it suffices to show that $\mu(C) = m(C)$ for all closed subsets C of X . Let $\epsilon > 0$, by regularity of the measure m there exists an open set U_ϵ such that $C \subseteq U_\epsilon$ and $m(U_\epsilon \setminus C) < \epsilon$. Define $f \in C(X)$ as follows

$$f(x) = \frac{d(x, X \setminus U_\epsilon)}{d(x, X \setminus U_\epsilon) + d(x, C)}.$$

Notice that $1_C \leq f \leq 1_{U_\epsilon}$, thus

$$\mu(C) \leq \int_X f \, d\mu = \int_X f \, dm \leq m(U_\epsilon) \leq m(C) + \epsilon.$$

Using a similar argument, one can show that $m(C) \leq \mu(C) + \epsilon$. Therefore, $\mu(C) = m(C)$ for all closed sets, and hence for all Borel sets. \square

This allows us to define a metric structure on $M(X)$ as follows. A sequence $\{\mu_n\}$ in $M(X)$ converges to $\mu \in M(X)$ if and only if

$$\lim_{n \rightarrow \infty} \int_X f(x) \, d\mu_n(x) = \int_X f(x) \, d\mu(x)$$

for all $f \in C(X)$. We will show that under this notion of convergence the space $M(X)$ is compact, but first we need The Riesz Representation Representation Theorem.

Theorem 6.1.3 (The Riesz Representation Theorem) *Let X be a compact metric space and $J : C(X) \rightarrow \mathbb{C}$ a continuous linear map such that J is a positive operator and $J(1) = 1$. Then there exists a $\mu \in M(X)$ such that $J(f) = \int_X f(x) \, d\mu(x)$.*

Theorem 6.1.4 *The space $M(X)$ is compact.*

Idea of proof. Let $\{\mu_n\}$ be a sequence in $M(X)$. Choose a countable dense subset of $\{f_n\}$ of $C(X)$. The sequence $\{\int_X f_1 d\mu_n\}$ is a bounded sequence of complex numbers, hence has a convergent subsequence $\{\int_X f_1 d\mu_n^{(1)}\}$. Now, the sequence $\{\int_X f_2 d\mu_n^{(1)}\}$ is bounded, and hence has a convergent subsequence $\{\int_X f_2 d\mu_n^{(2)}\}$. Notice that $\{\int_X f_1 d\mu_n^{(2)}\}$ is also convergent. We continue in this manner, to get for each i a subsequence $\{\mu_n^{(i)}\}$ of $\{\mu_n\}$ such that for all $j \leq i$, $\{\mu_n^{(i)}\}$ is a subsequence of $\{\mu_n^{(j)}\}$ and $\{\int_X f_j d\mu_n^{(i)}\}$ converges. Consider the diagonal sequence $\{\mu_n^{(n)}\}$, then $\{\int_X f_j d\mu_n^{(n)}\}$ converges for all j , and hence $\{\int_X f d\mu_n^{(n)}\}$ converges for all $f \in C(X)$. Now define $J : C(X) \rightarrow \mathbb{C}$ by $J(f) = \lim_{n \rightarrow \infty} \{\int_X f d\mu_n^{(n)}\}$. Then, J is linear, continuous ($|J(f)| \leq \sup_{x \in X} |f(x)|$), positive and $J(1) = 1$. Thus, by Riesz Representation Theorem, there exists a $\mu \in M(X)$ such that $J(f) = \lim_{n \rightarrow \infty} \{\int_X f d\mu_n^{(n)}\} = \int_X f d\mu$. Therefore, $\lim_{n \rightarrow \infty} \mu_n^{(n)} = \mu$, and $M(X)$ is compact. \square

Let $T : X \rightarrow X$ be a continuous transformation. Since \mathcal{B} is generated by the open sets, then T is measurable with respect to \mathcal{B} . Furthermore, T induces in a natural way, an operator $\bar{T} : M(X) \rightarrow M(X)$ given by

$$(\bar{T}\mu)(A) = \mu(T^{-1}A)$$

for all $A \in \mathcal{B}$. Then $\bar{T}^i \mu(A) = \mu(T^{-i}A)$. Using a standard argument, one can easily show that

$$\int_X f(x) d(\bar{T}\mu)(x) = \int_X f(Tx) d\mu(x)$$

for all continuous functions f on X . Note that T is measure preserving with respect to $\mu \in M(X)$ if and only if $\bar{T}\mu = \mu$. Equivalently, μ is measure preserving if and only if

$$\int_X f(x) d\mu(x) = \int_X f(Tx) d\mu(x)$$

for all continuous functions f on X . Let

$$M(X, T) = \{\mu \in M(X) : \bar{T}\mu = \mu\}$$

be the collection of all probability measures under which T is measure preserving.

Theorem 6.1.5 Let $T : X \rightarrow X$ be continuous, and $\{\sigma_n\}$ a sequence in $M(X)$. Define a sequence $\{\mu_n\}$ in $M(X)$ by

$$\mu_n = \frac{1}{n} \sum_{i=0}^{n-1} T^i \sigma_n.$$

Then, any limit point μ of $\{\mu_n\}$ is a member of $M(X, T)$.

Proof. We need to show that for any continuous function f on X , one has $\int_X f(x) d\mu(x) = \int_X f(Tx) d\mu$. Since $M(X)$ is compact there exists a $\mu \in M(X)$ and a subsequence $\{\mu_{n_j}\}$ such that $\mu_{n_j} \rightarrow \mu$ in $M(X)$. Now for any f continuous, we have that

$$\begin{aligned} & \left| \int_X f(Tx) d\mu(x) - \int_X f(x) d\mu(x) \right| \\ &= \lim_{j \rightarrow \infty} \left| \int_X f(Tx) d\mu_{n_j}(x) - \int_X f(x) d\mu_{n_j}(x) \right| \\ &= \lim_{j \rightarrow \infty} \left| \frac{1}{n_j} \int_X \sum_{i=0}^{n_j-1} (f(T^{i+1}x) - f(T^i x)) d\sigma_{n_j}(x) \right| \\ &= \lim_{j \rightarrow \infty} \left| \frac{1}{n_j} \int_X (f(T^{n_j}x) - f(x)) d\sigma_{n_j}(x) \right| \\ &\leq \lim_{j \rightarrow \infty} \frac{2 \sup_{x \in X} |f(x)|}{n_j} = 0. \end{aligned}$$

Therefore $\mu \in M(X, T)$. □

Theorem 6.1.6 Let T be a continuous transformation on a compact metric space. Then,

- (i) $M(X, T)$ is a compact convex subset of $M(X)$.
- (ii) $\mu \in M(X, T)$ is an extreme point (i.e. μ cannot be written in a non-trivial way as a convex combination of elements of $M(X, T)$) if and only if T is ergodic with respect to μ .

Proof. (i) Clearly $M(X, T)$ is convex. Now let $\{\mu_n\}$ be a sequence in $M(X, T)$ converging to μ in $M(X)$. We need to show that $\mu \in M(X, T)$. Since T is

continuous, then for any continuous function f on X , the function $f \circ T$ is also continuous. Hence,

$$\begin{aligned} \int_X f(Tx) \, d\mu(x) &= \lim_{n \rightarrow \infty} \int_X f(Tx) \, d\mu_n(x) \\ &= \lim_{n \rightarrow \infty} \int_X f(x) \, d\mu_n(x) \\ &= \int_X f(x) \, d\mu(x). \end{aligned}$$

Therefore, T is measure preserving with respect to μ , and $\mu \in M(X, T)$.

(ii) Suppose T is ergodic with respect to μ , and assume that

$$\mu = p\mu_1 + (1 - p)\mu_2$$

for some $\mu_1, \mu_2 \in M(X, T)$, and some $0 < p \leq 1$. We will show that $\mu = \mu_1$. Notice that the measure μ_1 is absolutely continuous with respect to μ , and T is ergodic with respect to μ , hence by Theorem (2.1.2) we have $\mu_1 = \mu$.

Conversely, (we prove the contrapositive) suppose that T is not ergodic with respect to μ . Then there exists a measurable set E such that $T^{-1}E = E$, and $0 < \mu(E) < 1$. Define measures μ_1, μ_2 on X by

$$\mu_1(B) = \frac{\mu(B \cap E)}{\mu(E)} \text{ and } \mu_2(B) = \frac{\mu(B \cap (X \setminus E))}{\mu(X \setminus E)}.$$

Since E and $X \setminus E$ are T -invariant sets, then $\mu_1, \mu_2 \in M(X, T)$, and $\mu_1 \neq \mu_2$ since $\mu_1(E) = 1$ while $\mu_2(E) = 0$. Furthermore, for any measurable set B ,

$$\mu(B) = \mu(E)\mu_1(B) + (1 - \mu(E))\mu_2(B),$$

i.e., μ is a non-trivial convex combination of elements of $M(X, T)$. Thus, μ is not an extreme point of $M(X, T)$. \square

Since the Banach space $C(X)$ of all continuous functions on X (under the sup norm) is separable i.e. $C(X)$ has a countable dense subset, one gets the following strengthening of the Ergodic Theorem.

Theorem 6.1.7 *If $T : X \rightarrow X$ is continuous and $\mu \in M(X, T)$ is ergodic, then there exists a measurable set Y such that $\mu(Y) = 1$, and*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) = \int_X f(x) \, d\mu(x)$$

for all $x \in Y$, and $f \in C(X)$.

Proof. Choose a countable dense subset $\{f_k\}$ in $C(X)$. By the Ergodic Theorem, for each k there exists a subset X_k with $\mu(X_k) = 1$ and

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f_k(T^i x) = \int_X f_k(x) d\mu(x)$$

for all $x \in X_k$. Let $Y = \bigcap_{k=1}^{\infty} X_k$, then $\mu(Y) = 1$, and

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f_k(T^i x) = \int_X f_k(x) d\mu(x)$$

for all k and all $x \in Y$. Now, let $f \in C(X)$, then there exists a subsequence $\{f_{k_j}\}$ converging to f in the supremum norm, and hence is uniformly convergent. For any $x \in Y$, using uniform convergence and the dominated convergence theorem, one gets

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) &= \lim_{n \rightarrow \infty} \lim_{j \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f_{k_j}(T^i x) \\ &= \lim_{j \rightarrow \infty} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f_{k_j}(T^i x) \\ &= \lim_{j \rightarrow \infty} \int_X f_{k_j} d\mu = \int_X f d\mu. \end{aligned}$$

□

Theorem 6.1.8 *Let $T : X \rightarrow X$ be continuous, and $\mu \in M(X, T)$. Then T is ergodic with respect to μ if and only if*

$$\frac{1}{n} \sum_{i=0}^{n-1} \delta_{T^i x} \rightarrow \mu \quad a.e.$$

Proof. Suppose T is ergodic with respect to μ . Notice that for any $f \in C(X)$,

$$\int_X f(y) d(\delta_{T^i x})(y) = f(T^i x),$$

Hence by Theorem 6.1.7, there exists a measurable set Y with $\mu(Y) = 1$ such that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \int_X f(y) d(\delta_{T^i x})(y) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) = \int_X f(y) d\mu(y)$$

for all $x \in Y$, and $f \in C(X)$. Thus, $\frac{1}{n} \sum_{i=0}^{n-1} \delta_{T^i x} \rightarrow \mu$ for all $x \in Y$.

Conversely, suppose $\frac{1}{n} \sum_{i=0}^{n-1} \delta_{T^i x} \rightarrow \mu$ for all $x \in Y$, where $\mu(Y) = 1$. Then for any $f \in C(X)$ and any $g \in L^1(X, \mathcal{B}, \mu)$ one has

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x)g(x) = g(x) \int_X f(y) d\mu(y).$$

By the dominated convergence theorem

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \int_X f(T^i x)g(x) d\mu(x) = \int_X g(x)d\mu(x) \int_X f(y) d\mu(y).$$

Now, let $F, G \in L^2(X, \mathcal{B}, \mu)$. Then, $G \in L^1(X, \mathcal{B}, \mu)$ so that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \int_X f(T^i x)G(x) d\mu(x) = \int_X G(x)d\mu(x) \int_X f(y)d\mu(y)$$

for all $f \in C(X)$. Let $\epsilon > 0$, there exists $f \in C(X)$ such that $\|F - f\|_2 < \epsilon$ which implies that $|\int F d\mu - \int f d\mu| < \epsilon$. Furthermore, there exists N so that for $n \geq N$ one has

$$\left| \int_X \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x)G(x) d\mu(x) - \int_X G d\mu \int_X f d\mu \right| < \epsilon.$$

Thus, for $n \geq N$ one has

$$\begin{aligned}
& \left| \int_X \frac{1}{n} \sum_{i=0}^{n-1} F(T^i x) G(x) \, d\mu(x) - \int_X G \, d\mu \int_X F \, d\mu \right| \\
& \leq \int_X \frac{1}{n} \sum_{i=0}^{n-1} |F(T^i x) - f(T^i x)| |G(x)| \, d\mu(x) \\
& \quad + \left| \int_X \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) G(x) \, d\mu(x) - \int_X G \, d\mu \int_X f \, d\mu \right| \\
& \quad + \left| \int_X f \, d\mu \int_X G \, d\mu - \int_X F \, d\mu \int_X G \, d\mu \right| \\
& < \epsilon \|G\|_2 + \epsilon + \epsilon \|G\|_2.
\end{aligned}$$

Thus,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} \int_X F(T^i x) G(x) \, d\mu(x) = \int_X G(x) \, d\mu(x) \int_X F(y) \, d\mu(y)$$

for all $F, G \in L^2(X, \mathcal{B}, \mu)$ and $x \in Y$. Taking F and G to be indicator functions, one gets that T is ergodic. □

Exercise 6.1.1 Let X be a compact metric space and $T : X \rightarrow X$ be a continuous homeomorphism. Let $x \in X$ be periodic point of T of period n , i.e. $T^n x = x$ and $T^j x \neq x$ for $j < n$. Show that if $\mu \in M(X, T)$ is ergodic and $\mu(\{x\}) > 0$, then $\mu = \frac{1}{n} \sum_{i=0}^{n-1} \delta_{T^i x}$.

6.2 Unique Ergodicity and Uniform Distribution

A continuous transformation $T : X \rightarrow X$ on a compact metric space is *uniquely ergodic* if there is only one T -invariant probability measure μ on X . In this case, $M(X, T) = \{\mu\}$, and μ is necessarily ergodic, since μ is an

extreme point of $M(X, T)$. Recall that if $\nu \in M(X, T)$ is ergodic, then there exists a measurable subset Y such that $\nu(Y) = 1$ and

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) = \int_X f(y) d\nu(y)$$

for all $x \in Y$ and all $f \in C(X)$. When T is uniquely ergodic we will see that we have a much stronger result.

Theorem 6.2.1 *Let $T : X \rightarrow X$ be a continuous transformation on a compact metric space X . Then the following are equivalent:*

- (i) *For every $f \in C(X)$, the sequence $\left\{ \frac{1}{n} \sum_{j=0}^{n-1} f(T^j x) \right\}$ converges uniformly to a constant.*
- (ii) *For every $f \in C(X)$, the sequence $\left\{ \frac{1}{n} \sum_{j=0}^{n-1} f(T^j x) \right\}$ converges pointwise to a constant.*
- (iii) *There exists a $\mu \in M(X, T)$ such that for every $f \in C(X)$ and all $x \in X$.*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) = \int_X f(y) d\mu(y).$$

- (iv) *T is uniquely ergodic.*

Proof. (i) \Rightarrow (ii) is immediate.

(ii) \Rightarrow (iii) Define $L : C(X) \rightarrow \mathbb{C}$ by

$$L(f) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x).$$

By assumption $L(f)$ is independent of x , hence L is well defined. It is easy to see that L is linear, continuous ($|L(f)| \leq \sup_{x \in X} |f(x)|$), positive and $L(1) = 1$. Thus, by the Riesz Representation Theorem there exists a probability measure $\mu \in M(X)$ such that

$$L(f) = \int_X f(x) d\mu(x)$$

for all $f \in C(X)$. But

$$L(f \circ T) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^{i+1}x) = L(f).$$

Hence,

$$\int_X f(Tx) \, d\mu(x) = \int_X f(x) \, d\mu(x).$$

Thus, $\mu \in M(X, T)$, and for every $f \in C(X)$,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) = \int_X f(x) \, d\mu(x)$$

for all $x \in X$.

(iii) \Rightarrow (iv) Suppose $\mu \in M(X, T)$ is such that for every $f \in C(X)$,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) = \int_X f(x) \, d\mu(x)$$

for all $x \in X$. Assume $\nu \in M(X, T)$, we will show that $\mu = \nu$. For any $f \in C(X)$, since the sequence $\{\frac{1}{n} \sum_{j=0}^{n-1} f(T^j x)\}$ converges pointwise to the constant function $\int_X f(x) \, d\mu(x)$, and since each term of the sequence is bounded in absolute value by the constant $\sup_{x \in X} |f(x)|$, it follows by the Dominated Convergence Theorem that

$$\lim_{n \rightarrow \infty} \int_X \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) \, d\nu(x) = \int_X \int_X f(x) \, d\mu(x) \, d\nu(y) = \int_X f(x) \, d\mu(x).$$

But for each n ,

$$\int_X \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) \, d\nu(x) = \int_X f(x) \, d\nu(x).$$

Thus, $\int_X f(x) \, d\mu(x) = \int_X f(x) \, d\nu(x)$, and $\mu = \nu$.

(iv) \Rightarrow (i) The proof is done by contradiction. Assume $M(X, T) = \{\mu\}$ and suppose $g \in C(X)$ is such that the sequence $\left\{\frac{1}{n} \sum_{j=0}^{n-1} g \circ T^j\right\}$ does not converge uniformly on X . Then there exists $\epsilon > 0$ such that for each N there exists $n > N$ and there exists $x_n \in X$ such that

$$\left| \frac{1}{n} \sum_{j=0}^{n-1} g(T^j x_n) - \int_X g \, d\mu \right| \geq \epsilon.$$

Let

$$\mu_n = \frac{1}{n} \sum_{j=0}^{n-1} \delta_{T^j x_n} = \frac{1}{n} \sum_{j=0}^{n-1} T^j \delta_{x_n}.$$

Then,

$$\left| \int_X g \, d\mu_n - \int_X g \, d\mu \right| \geq \epsilon.$$

Since $M(X)$ is compact, there exists a subsequence μ_{n_i} converging to $\nu \in M(X)$. Hence,

$$\left| \int_X g \, d\nu - \int_X g \, d\mu \right| \geq \epsilon.$$

By Theorem (6.1.5), $\nu \in M(X, T)$ and by unique ergodicity $\mu = \nu$, which is a contradiction. \square

Example 6.2.1 If T_θ is an irrational rotation, then T_θ is uniquely ergodic. This is a consequence of the above theorem as well as Weyl's Theorem on uniform distribution: for any Riemann integrable function f on $[0, 1)$, and any $x \in [0, 1)$, one has

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x + i\theta - [x + i\theta]) = \int_X f(y) \, dy.$$

We now give another proof using Theorem 6.2.1(ii). Suppose θ is irrational.

Note that the set of trigonometric polynomials $p(x) = \sum_{n=-N}^N c_n e^{2\pi i n x}$ are dense in $C([0, 1))$ (under the supremum norm), so it is enough to consider

continuous functions of the form $f(x) = e^{2\pi ikx}$ for some fixed k . Now for any x and any n ,

$$\frac{1}{n} \sum_{j=0}^{n-1} f(T_{\theta}^j x) = \frac{1}{n} \sum_{j=0}^{n-1} e^{2\pi ik(x+j\theta)} = \frac{1}{n} e^{2\pi ikx} \sum_{j=0}^{n-1} (e^{2\pi ik\theta})^j.$$

Thus,

$$\frac{1}{n} \sum_{j=0}^{n-1} f(T_{\theta}^j x) = \begin{cases} 1 & \text{if } k = 0, \\ \frac{e^{2\pi ikx}}{n} \cdot \frac{e^{2\pi ikn\theta} - 1}{e^{2\pi ik\theta} - 1} & \text{if } k \neq 0. \end{cases}$$

Taking limits we have

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^{n-1} f(T_{\theta}^j x) = \begin{cases} 1 & \text{if } k = 0, \\ 0 & \text{if } k \neq 0, \end{cases}$$

This shows that T_{θ} is uniquely ergodic.

Example 6.2.2 Consider the sequence of *first digits*

$$\{1, 2, 4, 8, 1, 3, 6, 1, \dots\}$$

obtained by writing the first decimal digit of each term in the sequence

$$\{2^n : n \geq 0\} = \{1, 2, 4, 8, 16, 32, 64, 128, \dots\}.$$

For each $k = 1, 2, \dots, 9$, let $p_k(n)$ be the number of times the digit k appears in the first n terms of the *first digit* sequence. The asymptotic relative frequency of the digit k is then $\lim_{n \rightarrow \infty} \frac{p_k(n)}{n}$. We want to identify this limit for each $k \in \{1, 2, \dots, 9\}$. To do this, let $\theta = \log_{10} 2$, then θ is irrational. For $k = 1, 2, \dots, 9$, let $J_k = [\log_{10} k, \log_{10}(k+1))$. By unique ergodicity of T_{θ} , we have for each $k = 1, 2, \dots, 9$,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_{J_k}(T_{\theta}^i(0)) = \lambda(J_k) = \log_{10} \left(\frac{k+1}{k} \right).$$

Returning to our original problem, notice that the first digit of 2^i is k if and only if

$$k \cdot 10^r \leq 2^i < (k+1) \cdot 10^r$$

for some $r \geq 0$. In this case,

$$r + \log_{10} k \leq i \log_{10} 2 = i\theta < r + \log_{10}(k + 1).$$

This shows that $r = \lfloor i\theta \rfloor$, and

$$\log_{10} k \leq i\theta - \lfloor i\theta \rfloor < \log_{10}(k + 1).$$

But $T_\theta^i(0) = i\theta - \lfloor i\theta \rfloor$, so that $T_\theta^i(0) \in J_k$. Summarizing, we see that the first digit of 2^i is k if and only if $T_\theta^i(0) \in J_k$. Thus,

$$\lim_{n \rightarrow \infty} \frac{p_k(n)}{n} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} 1_{J_k}(T_\theta^i(0)) = \log_{10} \left(\frac{k+1}{k} \right).$$

Exercise 6.2.1 Let X be a compact metric space, and \mathcal{B} the Borel σ -algebra on X . Let $T : X \rightarrow X$ be a continuous transformation. Let $N \geq 1$ and $x \in X$.

- (a) Show that $T^N x = x$ if and only if $\frac{1}{N} \sum_{i=0}^{N-1} \delta_{T^i x} \in M(X, T)$. (δ_y is the Dirac measure concentrated at the point y .)
- (b) Suppose $X = \{1, 2, \dots, N\}$ and $Ti = i + 1 \pmod{N}$. Show that T is uniquely ergodic. Determine the unique ergodic measure.

Exercise 6.2.2 Let X be a compact metric space, \mathcal{B} the Borel σ -algebra on X and $T : X \rightarrow X$ a uniquely ergodic continuous transformation. Let μ be the unique ergodic measure, and assume that $\mu(G) > 0$ for all non-empty open sets $G \subseteq X$.

- (a) Show that for each non-empty open subset G of X there exists a continuous function $f \in C(X)$, and a closed subset F of G of positive μ measure such that $f(x) = 1$ for $x \in F$, $f(x) > 0$ for $x \in G$ and $f(x) = 0$ for $x \in X \setminus G$.
- (b) Show that for each $x \in X$ and for every non-empty open set $G \subseteq X$, there exists $n \geq 0$ such that $T^n x \in G$. Conclude that $\{T^n x : n \geq 0\}$ is dense in X .

Bibliography

- [Abr] Abramov, L.M. – *Entropy of induced automorphisms*, Dokl. Akad. Nauk SSSR **128** (1959), 647-650.
- [Ada] Adams, W.W. – *On a relationship between the convergents of the nearest integer and regular continued fractions*, Math. Comp. **33** (1979), 1321-1331.
- [Bil] Billingsley, P. – *Ergodic Theory and Information*, John Wiley and Sons, 1965.
- [Bor] Borel, É. – *Contribution a l'analyse arithmétiques du continu*, J. Math. Pures et Appl. (5) **9** (1903), 329-375.
- [BJW] Bosma, W., H. Jager and F. Wiedijk – *Some metrical observations on the approximation by continued fractions*, *ibid.* **45** (1983), 281-270.
- [BK] Bosma, W. and C. Kraaikamp – *Metrical theory for optimal continued fractions*, J. Number Theory **34** (1990), 251-270.
- [BP] de Bruijn, N.G. and K.A. Post – *A remark on uniformly distributed sequences and Riemann integrability*, Indag. Math. **30**, (1968), 149-150.
- [BS] Brin, M., and Stuck, G., *Introduction to Dynamical Systems*, Cambridge University Press, 2002.
- [CFS] Cornfeld, I. P., S. V. Fomin and Ya. G. Sinai – *Ergodic Theory*, Springer-Verlag, 1981.
- [DF] Dajani, K., and Fieldsteel, A. *Equipartition of Interval Partitions and an Application to Number Theory*, Proc. AMS Vol 129, **12**, (2001), 3453 - 3460.

- [DK] Dajani, K., and Kraaikamp, C., *Ergodic theory of numbers*, Carus Mathematical Monographs, 29. Mathematical Association of America, Washington, DC 2002.
- [DKS] Dajani, K.; Kraaikamp, C.; Solomyak, B. – *The natural extension of the β -transformation*, Acta Math. Hungar. **73** (1996), no. 1-2, 97–109.
- [De] Devaney, R.L., An Introduction to Chaotic Dynamical Systems, Second Edition, Addison-Wesley Publishing Company, Inc., 1989.
- [Doe] Doebelin, W. – *Remarques sur la théorie métrique des fractions continues*, Comp. Math. **7**, (1940), 353-371.
- [Ge] Gel'fond, A.O. – *A common property of number systems*, Izv. Akad. Nauk SSSR. Ser. Mat. **23** (1959) 809–814.
- [G] Gleick, J., Chaos: Making a New Science, Vintage, 1998.
- [H] Halmos, P.R. – *Measure Theory*, D. Van Nostrand Company, Inc., New York, N.Y., 1950.
- [Ho] Hoppensteadt, F.C., Analysis and Simulation of Chaotic Systems, Springer-Verlag New York, Inc., 1993.
- [Hu] W. Hurewicz, *Ergodic theorem without invariant measure*, Annals of Math., **45** (1944), 192–206.
- [Jag1] Jager, H. – *On the speed of convergence of the nearest integer continued fraction*, Math. Comp. **39**, (1982), 555-558.
- [IK] Iosifescu, I., and Kraaikamp, C., *Metrical theory of continued fractions*. Mathematics and its Applications, 547. Kluwer Academic Publishers, Dordrecht, 2002.
- [Jag2] Jager, H. – *The distribution of certain sequences connected with the continued fraction*, Indag. Math. **48**, (1986), 61-69.
- [JK] Jager, H. and C. Kraaikamp – *On the approximation by continued fractions*, Indag. Math., **51** (1989), 289-307.
- [Knu2] Knuth, D.E. – *The distribution of continued fraction approximations*, J. Number Theory **19**, (1984), 443-448.

- [KN] Kuipers, L. and H. Niederreiter – *Uniform Distribution of Sequences*, John Wiley and Sons, 1974.
- [KK] Kamae, T., and Keane, M.S., *A simple proof of the ratio ergodic theorem*, Osaka J. Math. 34 (1997), no. 3, 653–657.
- [Knu] Knuth, D.E. – *The distribution of continued fraction approximations*, J. Number Theory **19**, (1984), 443-448.
- [KT] Kingman and Taylor, *Introduction to measure and probability*, Cambridge Press, 1966.
- [L] Lüroth, J. – *Ueber eine eindeutige Entwicklung von Zahlen in eine unendliche Reihe*, Math. Annalen **21** (1883), 411-423.
- [Le] Lévy, P. – *Sur le loi de probabilité dont dependent les quotients complets wet incompletes d'une fraction continue*, Bull. Soc. Math. de France **57**, (1929), 178-194.
- [Min] Minkowski, H. – *Über die Annäherung an eine reelle Grösse durch rationale Zahlen*, Math. Ann. **54**, (1901), 91-124.
- [M] Misiurewicz, M., *A short proof of the variational principle for a \mathbb{Z}_+^n action on a compact space*, Asterisque, 40, , 1976.
- [Mu] Munkres, J.R., *Topology*, Second Edition, Prentice Hall, Inc., 2000.
- [Nak] Nakada, H. – *Metrical theory for a class of continued fraction transformations and their natural extensions*, Tokyo J. Math. **4**, (1981), 399-426.
- [Obr] Obrechhoff, N. – *Sur l'approximation des nombres irrationnels par des nombres rationels*, C. R. Acad. Bulgare Sci. **3** (1), (1951), 1-4.
- [Pa] Parry, W., *Topics in Ergodic Theory*, Reprint of the 1981 original. Cambridge Tracts in Mathematics, 75. Cambridge University Press, Cambridge, 2004.
- [Pa2] Parry, W. – *On the β -expansion of real numbers*, Acta Math. Acad. Sci. Hungary **11** (1960), 401-416.

- [Per] Perron, O. – *Die Lehre von den Kettenbrüchen*, Band I, 3. verb. u. etw. Aufl., B. G. Teubner, Stuttgart.
- [P] Petersen, K., *Ergodic Theory*, Cambridge Studies in Advanced Mathematics, 2. Cambridge University Press, Cambridge, 1989.
- [Rie1] Rieger, G. J. – *Ein Gauss-Kusmin-Lévy Satz für Kettenbrüche nach nächsten Ganzen*, Manuscripta Math. **24**, (1978), 437-448.
- [Rie2] Rieger, G. J. – *Über die Länge von Kettenbrüchen mit ungeraden Teilnennern*, Abh. Braunschweig Wiss. Ges. **32**, (1981), 61-69.
– *On the metrical theory of the continued fractions with odd partial quotients*, in: *Topics in Classical Number Theory*, vol. II, Budapest 1981, 1371-1481 (Colloq. Math. Soc. János Bolyai, **34**, North-Holland 1984).
- [Ro] Rohlin, V.A. – *Exact endomorphisms of a Lebesgue space*, Izv. Akad. Naik SSSR, Ser. Mat., **24** (1960); English AMS translation, Series **2**, **39** (1969), 1-36.
- [RN] Ryll-Nardzewski, C. – *On the ergodic theorems (II) (Ergodic theory of continued fractions)*, Studia Math. **12**, (1951), 74-79.
- [S1] Schweiger, F. – *Continued fractions with odd and even partial quotients*, Arbeitsbericht Math. Inst. Univ. Salzburg **4**, (1982), 59-70.
– *On the approximation by continued fractions with odd and even partial quotients*, Arbeitsbericht Math. Inst. Univ. Salzburg **1-2**, (1984), 105-114.
- [S2] Schweiger, F. – *Ergodic Theory of Fibered Systems and Metric Number Theory*, Clarendon Press, Oxford 1995.
- [Sen] Sendov, B. – *Der Vahlensche Satz über die singulären Kettenbrüche und die Kettenbrüche nach nächsten Ganzen*, Ann. Univ. Sofia, Fac. Sci. Math. Livre 1 Math. **54**, (1959/60), 251-258.
- [St] Steuding, J. – *Diophantine analysis. Discrete Mathematics and its Applications*. Chapman & Hall/CRC, Boca Raton, FL, 2005.
- [Tie] Tietze, H. – *Über die raschesten Kettenbruchentwicklung reeller Zahlen*, Monatsh. Math. Phys. **24**, (1913), 209-241.

- [W] Walters, P., *An Introduction to Ergodic Theory*, Graduate Texts in Mathematics, 79. Springer-Verlag, New York-Berlin, 1982.