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Generalizations of Wiener-Wintner ergodic theorem

Praca magisterska
na kierunku MATEMATYKA

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Oświadczenie kierującego prac

Potwierdzam, że niniejsza praca została przygotowana pod moim kierunkiem i kwalifikuje się do przedstawienia jej w postępowaniu o nadanie tytułu zawodowego.

Data

Podpis kierującego prac

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Oświadczam ponadto, że niniejsza wersja pracy jest identyczna z załączoną wersją elektroniczną.

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Abstract

The Wiener-Wintner ergodic theorem (3.1) states that for a given ergodic dynamical system (X, \mathcal{A}, μ, T) and fixed $f \in L^1(\mu)$ there exists a set of full measure X_f with $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x)$ exists for every $x \in X_f$ and λ from the unit circle \mathbb{T} on the complex plane. We state and prove several generalizations of this theorem. Bourgain's theorem (3.2) establishes uniform convergence with respect to $\lambda \in \mathbb{T}$ and Robinson's theorem (3.3) gives uniform convergence with respect to $x \in X$ in the case of uniquely ergodic topological systems (under suitable additional assumptions). Schreiber's theorem (5.1) gives condition equivalent to the mean ergodicity of distorted semigroup $\{\chi(g)S_g : g \in G\}$ for a given uniquely ergodic Markovian representation \mathcal{S} of an amenable semitopological semigroup G on $C(X)$ and continuous character $\chi : G \rightarrow \mathbb{T}$. This can be seen as a natural extension of Robinson's theorem to the abstract semigroup case. We finish the thesis with generalization of Schreiber's theorem to the case of mean ergodic semigroup \mathcal{S} (5.2) with finite dimensional fixed space $Fix(\mathcal{S})$ (5.3), giving simultaneously more elementary proof. We present also necessary framework from measure theory, functional analysis, spectral theory for isometries on Hilbert spaces, Markov operators and ergodic theory for deterministic transformations and operator semigroups.

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37A Ergodic theory

37A30 Ergodic theorems, spectral theory, Markov operators

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Rozszerzenia twierdzenia ergodycznego Wienera-Wintnera

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Introduction

Chapter 1

Preliminaries

In this chapter we introduce basic notations, concepts and theorems from measure theory, topology and functional analysis which will be used through the thesis. We omit most of the proofs.

By \mathbb{N} we will denote set of positive natural numbers, by \mathbb{N}_0 - set of natural numbers with zero, by \mathbb{Z} - set of integers, by \mathbb{R} - set of real numbers, by \mathbb{C} - set of complex numbers and by $\mathbb{T} = \mathbb{S}^1 = \{\lambda \in \mathbb{C} : |\lambda| = 1\}$ - circle on a complex plane (1-dimensional torus).

1.1. Measure theory

By (X, \mathcal{A}, μ) we will denote a measure space, where X is a nonempty set, \mathcal{A} is a σ -field of subsets of X and μ is a measure on the measurable space (X, \mathcal{A}) . By term 'measure' we will always mean a σ -additive, non-negative measure. When we will wish measure μ to have complex values we will introduce it as a 'complex measure μ '. Sets $A \in \mathcal{A}$ are called measurable sets. Measure μ is called finite if $\mu(X) < \infty$ and σ -finite if there is a countable collection of measurable sets $\{A_n\}_{n=1}^{\infty}$ with $\mu(A_n) < \infty$ for each $n \in \mathbb{N}$, such that $X = \bigcup_{n=1}^{\infty} A_n$. If measure μ is σ -finite, then the sets $\{A_n\}_{n=1}^{\infty}$ can be taken to be pairwise disjoint. If $\mu(X) = 1$ then the measure μ is called a probability measure and (X, \mathcal{A}, μ) is called a probability space. For a finite measure μ , set $A \in \mathcal{A}$ is said to have a full measure if $\mu(A) = \mu(X)$. We will often use the following simple

Fact 1.1 *Let (X, \mathcal{A}, μ) be a measure space with a finite measure and take measurable sets $A_n, n \in \mathbb{N}$. If each A_n has a full measure, then their intesection $\bigcap_{n=1}^{\infty} A_n$ also has a full measure.*

Definition 1.1 Let (X, \mathcal{A}) and (Y, \mathcal{C}) be measurable spaces. A map $T : X \rightarrow Y$ is called a **measurable map** if it satisfies $T^{-1}(C) \in \mathcal{A}$ for all $C \in \mathcal{C}$.

Definition 1.2 Let (X, \mathcal{A}, μ) be a measure space. An element $x \in X$ is called an **atom** (of the measure μ) if $\mu(\{x\}) > 0$. The measure μ is called **continuous** if it has no atoms, i.e. $\bigvee_{x \in X} \mu(\{x\}) = 0$.

Remark Note that a finite measure μ can have only countably many atoms. To see that observe that for $\varepsilon > 0$ a set $A_\varepsilon := \{x \in X : \mu(\{x\}) > \varepsilon\}$ must have at most $\frac{\mu(X)}{\varepsilon}$ elements (otherwise

we would have $\mu(X) > \frac{\mu(X)}{\varepsilon} \cdot \varepsilon = \mu(X)$, hence must be finite. This shows that the set of atoms $A = \bigcup_{n=1}^{\infty} A_{\frac{1}{n}}$ must be countable. Also, there is $\sum_{x \in A} \mu(\{x\}) \leq \mu(X) < \infty$.

If X is a topological space, then by $\mathcal{B}(X)$ we will denote its Borel σ -field, i.e. the smallest σ -field containing all open subsets of X . Note that if X and Y are topological spaces and $T : X \rightarrow Y$ is continuous, then T is also measurable (with respect to Borel σ -fields on X and Y). Measure on a measurable space $(X, \mathcal{B}(X))$ is called a Borel measure. On spaces $\mathbb{R}^n, n \in \mathbb{N}$ (with a standard topology) there is a natural Borel measure, which is a unique measure m with property $m([a_1, b_1] \times [a_2, b_2] \times \dots \times [a_n, b_n]) = (b_1 - a_1)(b_2 - a_2) \dots (b_n - a_n)$ for $a_1 \leq b_1, a_2 \leq b_2, \dots, a_n \leq b_n$. This measure m is called a Lebesgue measure and it's σ -finite. On \mathbb{T} (with a standard topology) there exists a unique measure such that measure of any arc is its length. This measure is finite and its normalization will be also called a Lebesgue measure and will be denoted by m ($m(\mathbb{T}) = 1$ and length of arc A is equal to $2\pi m(A)$).

Let $T, S : X \rightarrow Y$ be measurable maps between measurable spaces (X, \mathcal{A}) and (Y, \mathcal{C}) . We will use abbreviations $\{T \in A\} := \{x \in X : Tx \in A\}$, $\{T = S\} := \{x \in X : Tx = Sx\}$ and $\{T \neq S\} := \{x \in X : Tx \neq Sx\}$. Note that these sets are measurable. We will say that some property holds for (μ) almost all $x \in X$, if there is a measurable set A such that this property holds for every $x \in A$ and $\mu(X \setminus A) = 0$. We will say that map T is equal to S μ almost everywhere (μ -a.e.) if $Tx = Sx$ for almost all $x \in X$, i.e. $\mu(T \neq S) := \mu(\{T \neq S\}) = 0$.

We say that function $f : X \rightarrow \mathbb{R}^n$ or $f : X \rightarrow \mathbb{C}^n$ is a Borel function (or simply a measurable function) if it is measurable with respect to the Borel σ -field $\mathcal{B}(\mathbb{R}^n)$ or $\mathcal{B}(\mathbb{C}^n)$, where we consider the standard topology on \mathbb{R}^n or \mathbb{C}^n . Let $(f_n)_{n \in \mathbb{N}}$ be a sequence of measurable complex-valued functions on X . We say that the sequence $(f_n)_{n \in \mathbb{N}}$ converges μ almost everywhere to a measurable function $f : X \rightarrow \mathbb{C}$ if $\lim_{n \rightarrow \infty} f_n(x) = f(x)$ for almost all $x \in X$. The following important theorem says when a.e. convergence implies convergence of integrals.

Theorem 1.1 (Lebesgue Dominated Convergence Theorem)

Let (X, \mathcal{A}, μ) be a measure space and let f, f_1, f_2, \dots be a sequence of measurable, complex-valued functions on X with $f_n \xrightarrow{n \rightarrow \infty} f$ μ -a.e. Suppose further, that there is a measurable function $g : X \rightarrow [0, \infty)$ with $|f_n| \leq g$ μ -a.e. for every $n \in \mathbb{N}$ and $\int_X g d\mu < \infty$. Then functions f, f_1, f_2, \dots are integrable and

$$\lim_{n \rightarrow \infty} \int_X f_n d\mu = \int_X f d\mu.$$

Given two measure spaces (X, \mathcal{A}, μ) and (Y, \mathcal{C}, ν) there is a natural measurable structure on the product $X \times Y$. We define product σ -field $\mathcal{A} \otimes \mathcal{C}$ of the subsets of $X \times Y$ as the smallest σ -field containing all measurable rectangles $A \times C, A \in \mathcal{A}, C \in \mathcal{C}$, i.e. $\mathcal{A} \otimes \mathcal{C} := \sigma(\{A \times C : A \in \mathcal{A}, C \in \mathcal{C}\})$. Moreover, if both measures μ and ν are σ -finite, then there exists unique measure $\mu \otimes \nu$ on the measurable space $(X \times Y, \mathcal{A} \otimes \mathcal{C})$ with the property $\mu \otimes \nu(A \times C) = \mu(A)\nu(C)$ for all $A \in \mathcal{A}, C \in \mathcal{C}$. The measure $\mu \otimes \nu$ is called a product measure. Fubini's theorem establishes connection between integral with respect to the product measure and iterated integrals with respect to the measures μ and ν separately.

Theorem 1.2 (Fubini's Theorem)

Let (X, \mathcal{A}, μ) and (Y, \mathcal{C}, ν) be measure spaces with σ -finite measures. Let $f : X \times Y \rightarrow \mathbb{C}$ be

measurable with respect to the product σ -field $\mathcal{A} \otimes \mathcal{C}$ and suppose that at least one of the following integrals is finite:

$$\int_{X \times Y} |f| d\mu \otimes \nu, \int_X \left(\int_Y |f(x, y)| d\nu(y) \right) d\mu(x), \int_Y \left(\int_X |f(x, y)| d\mu(x) \right) d\nu(y).$$

Then for μ -almost all $x \in X$ the function $f(x, \cdot) : Y \rightarrow \mathbb{C}$ is ν -finitely integrable and for ν -almost all $y \in Y$ the function $f(\cdot, y) : X \rightarrow \mathbb{C}$ is μ -finitely integrable. Moreover function $X \ni x \mapsto \int_Y f(x, y) d\nu(y) \in \mathbb{C}$ is μ -finitely integrable and function $Y \ni y \mapsto \int_X f(x, y) d\mu(x) \in \mathbb{C}$ is ν -finitely integrable. The following equality holds:

$$\int_{X \times Y} f d\mu \otimes \nu = \int_X \left(\int_Y f(x, y) d\nu(y) \right) d\mu(x) = \int_Y \left(\int_X f(x, y) d\mu(x) \right) d\nu(y).$$

Remark Note that functions $X \ni x \mapsto \int_Y f(x, y) d\nu(y) \in \mathbb{C}$ and $Y \ni y \mapsto \int_X f(x, y) d\mu(x) \in \mathbb{C}$ may not be defined properly for every $x \in X$ and $y \in Y$, although they are defined μ - and ν -almost everywhere, which is enough to define properly their integrals.

There is also a version of Fubini's Theorem for non-negative functions. In this case the integrals do not need to be finite.

Theorem 1.3 (Fubini's Theorem for non-negative functions)

Let (X, \mathcal{A}, μ) and (Y, \mathcal{C}, ν) be measure spaces with σ -finite measures. Let $f : X \times Y \rightarrow [0, \infty)$ be measurable with respect to the product σ -field $\mathcal{A} \otimes \mathcal{C}$. Then function $X \ni x \mapsto \int_Y f(x, y) d\nu(y) \in \mathbb{C}$ is \mathcal{A} -measurable and function $Y \ni y \mapsto \int_X f(x, y) d\mu(x) \in \mathbb{C}$ is \mathcal{C} -measurable and the following equality holds:

$$\int_{X \times Y} f d\mu \otimes \nu = \int_X \left(\int_Y f(x, y) d\nu(y) \right) d\mu(x) = \int_Y \left(\int_X f(x, y) d\mu(x) \right) d\nu(y).$$

Remark Note that these integrals may be infinite and if at least one of them is infinite, then all of them are.

Definition 1.3 Let X be a compact topological space and let μ be a finite, non-negative Borel measure on X . We say that μ is **regular** if

- (i) for every $A \in \mathcal{B}(X)$ there is $\mu(A) = \inf\{\mu(U) : A \subset U, U \text{ is open}\}$,
- (ii) for every $A \in \mathcal{B}(X)$ there is $\mu(A) = \sup\{\mu(K) : K \subset A, K \text{ is compact}\}$.

Definition 1.4 Let μ be a complex measure on a measurable space (X, \mathcal{A}) . We define its total variation measure $|\mu|$ as a non-negative measure on (X, \mathcal{A}) given by

$$|\mu|(A) := \sup\left\{ \sum_{j=1}^{\infty} |\mu(E_j)| : E_1, E_2, \dots \text{ are pairwise disjoint and } \bigcup_{j=1}^{\infty} E_j = A \right\}.$$

$|\mu|$ is always a finite measure (note that a complex measure, unlike the non-negative or signed measure, cannot attain values $+\infty$ or $-\infty$ by definition). For every $A \in \mathcal{A}$ there is $|\mu(A)| \leq |\mu|(A)$ and $|\mu(A)| = |\mu|(A)$ does not hold in general. We say that complex Borel measure is regular if its total variation measure is regular.

1.2. Functional analysis

We will always assume that vector spaces are taken over field \mathbb{C} . By $\|\cdot\|$ we will denote a norm of a normed space. We will give now standard examples of Banach spaces (i.e. complete normed space) with their properties which will be useful for us later.

Example 1.1 (\mathcal{L}^p and L^p spaces)

Let (X, \mathcal{A}, μ) be a measure space. For $1 \leq p < \infty$ consider the vector space

$$\mathcal{L}^p(X, \mathcal{A}, \mu) := \left\{ f : X \rightarrow \mathbb{C}; f \text{ is measurable and } \int_X |f|^p d\mu < \infty \right\}.$$

Define an equivalence relation \sim on $\mathcal{L}^p(X, \mathcal{A}, \mu)$ by $f \sim g$ if $f = g$ μ a.e. Let

$$L^p(X, \mathcal{A}, \mu) := \mathcal{L}^p(X, \mathcal{A}, \mu) / \sim.$$

Space $L^p(X, \mathcal{A}, \mu)$ endowed with the norm $\|f\|_{L^p(X, \mathcal{A}, \mu)} := \left(\int_X |f|^p d\mu \right)^{\frac{1}{p}}$ become Banach spaces.

Usually we will abbreviate $L^p(X, \mathcal{A}, \mu)$ to $L^p(\mu)$ or L^p and $\|\cdot\|_{L^p(X, \mathcal{A}, \mu)}$ to $\|\cdot\|_{L^p(\mu)}$ or $\|\cdot\|_p$.

Proposition 1.1

Let (X, \mathcal{A}, μ) be a measure space with a finite measure. Then for $1 \leq p < q < \infty$ we have $\mathcal{L}^q(\mu) \subset \mathcal{L}^p(\mu)$ and $L^q(\mu)$ is dense in $L^p(\mu)$ (in L^p norm).

Example 1.2 (Space $C(X)$)

Let X be a compact topological space (we assume that compact spaces are Hausdorff by definition). Denote by $C(X)$ set of all complex valued continuous functions on X . $C(X)$ is a Banach space with norm $\|f\|_{\sup} := \sup_{x \in X} |f(x)|$, $f \in C(X)$. $C(X)$ is separable if and only if X is metrizable ([Eisner et al, thm. 4.7]). Suppose that there is a finite Borel nonnegative measure μ on X . Any function $f \in C(X)$ is bounded, hence integrable with any power $p \in [1, \infty)$, what means that $C(X) \subset \mathcal{L}^p(\mu)$ and $C(X)$ can be embedded into $L^p(X, \mathcal{B}(X), \mu)$. Therefore, space $C(X)$ can be naturally seen as a linear subspace of space $L^p(\mu)$ (with identification of functions equal μ a.e., so equivalence class $f \in L^2(\mu)$ is also in the subspace $C(X)$ if it has continuous representative).

Proposition 1.2

Let X be a compact topological space and μ be a finite nonnegative Borel measure on X . Then $C(X)$ is dense in $L^p(\mu)$ (in L^p norm) for every $p \in [1, \infty)$.

By $\langle \cdot, \cdot \rangle$ we will denote inner product on a inner product space. Inner product space is also a normed space with a norm $\|x\| := \sqrt{\langle x, x \rangle}$. If inner product space is complete with this norm, we call it a Hilbert space.

Example 1.3 (Space $L^2(\mu)$)

Let (X, \mathcal{A}, μ) be a measure space. The space $L^2(\mu)$ with inner product $\langle f, g \rangle := \int_X f \bar{g} d\mu$ is a Hilbert space. Note that the inner product norm coincides with norm $\|\cdot\|_{L^2(\mu)}$ from Example 1.2.

Proposition 1.3 (Cauchy–Schwarz inequality)

Let H be an inner product space. The following inequality holds for all $x, y \in H$:

$$|\langle x, y \rangle| \leq \|x\| \|y\|.$$

Definition 1.5 Let H be an inner product space. Two vectors $x, y \in H$ are said to be **orthogonal** if $\langle x, y \rangle = 0$ (we write $x \perp y$). For a set $H_0 \subset H$ its **orthogonal complement** is a set $H_0^\perp := \left\{ x \in H : \forall_{h \in H_0} \langle h, x \rangle = 0 \right\}$.

Remark If H_0 is a linear subspace of H , then H_0^\perp is a closed linear subspace of H .

Definition 1.6 Let E, F be normed spaces. A linear transformation $U : E \rightarrow F$ is called a **bounded linear operator** if there exists $M > 0$ such that $\forall_{x \in E} \|Ux\| \leq M\|x\|$. The constant $\|U\| := \sup_{\|x\| \leq 1} \|Ux\|$ is called a **operator norm** of U . If $\|U\| \leq 1$ then U is called a **contraction**. If $\forall_{x \in E} \|Ux\| = \|x\|$ then U is called an **isometry**. Note that an isometry is always a contraction.

Remark Linear operator $U : E \rightarrow F$ between normed spaces is continuous if and only if it's bounded. The space $L(E, F)$ of all bounded linear operators, endowed with the operator norm is a normed space. $L(E, F)$ is a Banach space if and only if F is a Banach space.

Definition 1.7 Let E be a normed space and let $U : E \rightarrow E$ be a (bounded) linear operator. Number $\lambda \in \mathbb{C}$ is called an **eigenvalue** if there is a vector $x \in E$, $x \neq 0$ such that $Ux = \lambda x$. Any such a vector x is called an **eigenvector** (associated with λ). The closed linear subspace $H_\lambda = \{x \in E : Ux = \lambda x\}$ is called an **eigenspace** (of λ). The set of all eigenvalues of operator U is called a **point spectrum** and is denoted by $\sigma_p(U)$.

Theorem 1.4 (Orthogonal Projection Theorem [Rudin1, lemma 12.4])

Let H_0 be a closed linear subspace of a Hilbert space H . Then

$$H = H_0 \oplus H_0^\perp,$$

i.e. for every $x \in H$ there are unique $x_0 \in H_0$, $x_1 \in H_0^\perp$ such that $x = x_0 + x_1$. Moreover, transformation $P : H \rightarrow H$ given by $P(x) = x_0$ is a bounded linear operator with $\|P\| \leq 1$ and $P \circ P = P$. Operator P is called an **orthogonal projection** on subspace H_0 .

For a normed space E we denote by E^* its dual space, i.e. the normed space of all continuous linear functionals $\Lambda : E \rightarrow \mathbb{C}$ with the operator norm. We use standard notation $\langle x, x^* \rangle := x^*(x)$ for $x \in E$, $x^* \in E^*$. We can consider weak topology on E , i.e. the coarsest topology such that each $\Lambda \in E^*$ is continuous and weak* topology on E^* , i.e. the coarsest topology such that for each $x \in E$ evaluation $E^* \ni \Lambda \mapsto \Lambda x \in \mathbb{C}$ is continuous. E with the weak topology and E^* with the weak* topology are locally convex topological vector spaces, although except the finite dimensional case they are never metrizable.

Theorem 1.5 (Banach-Alaouglu theorem [Rudin1, thm. 3.15])

Closed unit ball $\bar{B}(0, 1) \subset E^*$ in a dual space of a normed space E is weak* compact.

Theorem 1.6 ([Rudin1, thm. 3.16])

Let K be a weak* compact subset of a dual space E^* of a separable normed space E . Then the weak* topology is metrizable on K .

We say that $\Lambda \in E^*$ is a weak* limit of a sequence $(\Lambda_n)_{n \in \mathbb{N}}, \Lambda_n \in E^*$ if Λ_n converges to Λ in weak* topology, what is equivalent to the condition

$$\forall_{x \in E} \Lambda_n x \rightarrow \Lambda x.$$

We note this by $\Lambda_n \xrightarrow{*w} \Lambda$. Immediate corollary from above theorems is the following:

Corollary 1.1

Closed unit ball $\overline{B}(0, 1) \subset E^*$ in a dual space of a separable normed space E is weak* sequentially compact, i.e. for every sequence $(\Lambda_n)_{n \in \mathbb{N}}, \Lambda \in \overline{B}(0, 1)$ there is a subsequence $(\Lambda_{n_k})_{k \in \mathbb{N}}$ and $\Lambda \in \overline{B}(0, 1)$ with $\Lambda_{n_k} \xrightarrow{*w} \Lambda$.

Definition 1.8 Let V be a vector space and take $A \subset V$. Point $x \in A$ is said to be an **extremal point** of A if x is not a middle point of any interval with ends in A , i.e. if $x = \alpha y + (1 - \alpha)z$ for some $y, z \in A, \alpha \in (0, 1)$ then $y = z = x$. We denote set of all extreme points of A by $Ext(A)$.

Theorem 1.7 (Krein-Milman theorem [Rudin1, thm. 3.23])

Let E be a locally convex topological vector space. If K is a nonempty, compact and convex subset of E , then $K = \overline{co}(Ext(K))$, where $co(A)$ stands for a convex hull of a set $A \subset E$.

Corollary 1.2

Let K be a weak* compact subset of a dual space of a normed space E . Then $K = \overline{co}^{*w}(Ext(K))$, where \overline{A}^{*w} stands for closure of set $A \subset E^*$ in the weak* topology.

We will characterize now space $C(X)^*$ of bounded linear functionals on $C(X)$ for a compact space X .

Example 1.4 For a compact topological space X , by $\mathcal{M}(X)$ we will denote set of all (finite) complex-valued, regular Borel measures on X . $\mathcal{M}(X)$ is a Banach space with norm $\|\mu\| :=$

$\sup \left\{ \sum_{j=1}^n |\mu(A_j)| : n \in \mathbb{N}, A_1, \dots, A_n \in \mathcal{B}(X) \text{ are pairwise disjoint and } \bigcup_{j=1}^n A_j = X \right\} (= |\mu|(X))$. This norm is called a **variation norm**.

Definition 1.9 We say that functional $\Lambda \in C(X)^*$ is **positive** if $\Lambda f \geq 0$ for every $f \geq 0$ (i.e., $\forall_{x \in X} f(x) \in [0, \infty)$).

Theorem 1.8 (Riesz-Markov representation theorem)

Let X be a compact topological space. For every continuous linear functional $\Lambda \in C(X)^*$ there is a unique $\mu \in \mathcal{M}(X)$ with

$$\forall_{f \in C(X)} \Lambda f = \int_X f d\mu.$$

Moreover, for every $\mu \in \mathcal{M}(X)$ the above equality defines a unique continuous linear functional $\Lambda \in C(X)^*$ and there is $\|\Lambda\| = \|\mu\|$. Functional Λ is positive if and only if the corresponding measure μ is non-negative.

Remark Riesz-Markov theorem states that $\Phi : \mathcal{M}(X) \rightarrow C(X)^*$ given by $\Phi(\mu)(f) = \int_X f d\mu, f \in C(X)$ is a bijective isometry. It is straightforward to check that Φ is also linear. Hence we will identify Banach spaces $\mathcal{M}(X)$ and $C(X)^*$. We will use notation $\langle f, \mu \rangle = \int_X f d\mu$. This identification allows us also to consider weak* topology on $\mathcal{M}(X)$. Note that $\mu_n \xrightarrow{*w} \mu$ if and only if $\forall_{f \in C(X)} \int_X f d\mu_n \rightarrow \int_X f d\mu$.

Fact 1.2 Let X be a compact topological space. For every $\mu \in \mathcal{M}(X)$ there is

$$\langle f, |\mu| \rangle = \sup\{|\langle h, \mu \rangle| : h \in C(X) \text{ and } |h| \leq f\}.$$

For every $f \in C(X)$ there is

$$|\langle f, \mu \rangle| \leq \langle |f|, |\mu| \rangle.$$

Definition 1.10 Let E be a normed space. Denote by $L(E)$ set of all bounded linear operators $T : E \rightarrow E$. The **strong operator topology** (s.o.t) on $L(E)$ is a topology induced by a family of seminorms $\{p_x : x \in E\}$, where $p_x(T) := \|Tx\|$ for $x \in E, T \in L(E)$.

Remark Note that space $L(E)$ endowed with strong operator topology is a locally convex topological vector space. Net $(T_\alpha)_{\alpha \in \Lambda}$ of operators in $L(E)$ converges to $T \in L(E)$ in s.o.t if and only if

$$\forall_{x \in E} T_\alpha x \xrightarrow{\alpha} Tx,$$

so strong operator topology is a topology of pointwise convergence on $L(E)$.

Definition 1.11 Let E, F be normed space. For $T \in L(E, F)$ there exists unique operator $T' \in L(F^*, E^*)$ such that

$$T'y^*(x) = y^*(Tx) \quad (\langle Tx, y^* \rangle = \langle x, T'y^* \rangle) \quad \text{for all } x \in E, y^* \in F^*.$$

Operator T' is called an **adjoint (dual) operator** of T .

Definition 1.12 Let H be a Hilbert space and let $T : H \rightarrow H$ be a bounded linear operator. There exists unique operator $T^* : H \rightarrow H$ such that

$$\langle Tx, y \rangle = \langle x, T^*y \rangle \text{ for all } x, y \in H.$$

Operator T^* is called an **Hermitian adjoint operator** of T .

Definition 1.13 We say that Banach space E is reflexive if it is isometrically isomorphic to its second dual, i.e. $X \cong X^{**}$.

Banach-Alaoglu Theorem implies the following

Theorem 1.9

Banach space E is reflexive if and only if its closed unit ball is weakly compact.

Every Hilbert space is reflexive.

Theorem 1.10 (Krein-Smulian Theorem [Eisner et al, thm. G.7])

Let E be a Banach space. If set $A \subset E$ is relatively weakly compact then also $\text{co}A$ is relatively weakly compact.

We will also make use of the following topological theorem:

Theorem 1.11 (Urysohn's lemma [Rudin2, thm. 2.12])

Let X be a compact Hausdorff space and let F, G be compact and disjoint subsets of X . There exists continuous function $f : X \rightarrow [0, 1]$ with $f = 1$ on F and $f = 0$ on G .

1.3. Spectral theory for isometries

We will introduce now basic facts from spectral theory for isometries on Hilbert spaces.

Remark Let H be a complex inner product space. Then bounded linear operator $U : H \rightarrow H$ is an isometry if and only if $\forall_{x, y \in H} \langle Ux, Uy \rangle = \langle x, y \rangle$.

Definition 1.14 Sequence $(r_n)_{n \in \mathbb{Z}}$ of complex numbers is called **positive definite** if for every sequence $(a_n)_{n \in \mathbb{N}_0}$ of complex numbers and every $N \in \mathbb{N}_0$ we have $\sum_{n, m=0}^N r_{n-m} a_n \overline{a_m} \geq 0$.

Proposition 1.4

Let $U : H \rightarrow H$ be an isometry on Hilbert space H . For a vector $x \in H$ define $r_n := \langle U^n x, x \rangle$ for $n \geq 0$ and $r_n := \overline{r_{-n}} = \langle x, U^n x \rangle$ for $n < 0$. The sequence $(r_n)_{n \in \mathbb{Z}}$ is positive definite.

Proof: Note first that for $n \geq m$ we have $r_{n-m} = \langle U^{n-m} x, x \rangle = \langle U^n x, U^m x \rangle$ (since U is an isometry) and for $n < m$ we also have $r_{n-m} = \overline{r_{m-n}} = \overline{\langle U^{m-n} x, x \rangle} = \langle U^m x, U^n x \rangle = \langle U^n x, U^m x \rangle$. Now compute

$$\begin{aligned} \sum_{n, m=0}^N r_{n-m} a_n \overline{a_m} &= \sum_{n, m=0}^N \langle U^n x, U^m x \rangle a_n \overline{a_m} = \sum_{n, m=0}^N \langle a_n U^n x, a_m U^m x \rangle = \\ &= \sum_{n=0}^N \langle a_n U^n x, \sum_{m=0}^N a_m U^m x \rangle = \langle \sum_{n=0}^N a_n U^n x, \sum_{m=0}^N a_m U^m x \rangle = \left\| \sum_{n=0}^N a_n U^n x \right\|^2 \geq 0. \end{aligned} \tag{1.1}$$

□

Theorem 1.12 (Herglotz's theorem [Lemańczyk, thm. 2.3])

Let $(r_n)_{n \in \mathbb{Z}}$ be positive definite sequence. Then there exists a unique non-negative finite Borel measure σ on \mathbb{T} such that

$$r_n = \int_{\mathbb{T}} z^n d\sigma(z) \quad \text{for all } n \in \mathbb{Z}. \tag{1.2}$$

Conversly, for every non-negative finite Borel measure σ on \mathbb{T} , sequence r_n defined by (1.2) is positive definite.

Definition 1.15 Let σ be a non-negative finite Borel measure on \mathbb{T} . Then the quantity

$$\hat{\sigma}(n) := \int_{\mathbb{T}} z^n d\sigma(z), \quad n \in \mathbb{Z}$$

is called the **n-th Fourier coefficient** of the measure σ . Note that the sequence $\hat{\sigma}(n)$, $n \in \mathbb{Z}$ is positive definite and $\hat{\sigma}(-n) = \overline{\hat{\sigma}(n)}$ for every $n \in \mathbb{Z}$.

Corollary 1.3 (Spectral measure)

Let $U : H \rightarrow H$ be an isometry on Hilbert space H . For every vector $x \in H$ there exists unique non-negative finite Borel measure $\sigma_{x,U}$ on \mathbb{T} such that

$$\langle U^n x, x \rangle = \int_{\mathbb{T}} z^n d\sigma_{x,U}(z) \quad \text{and} \quad \langle x, U^n x \rangle = \int_{\mathbb{T}} z^{-n} d\sigma_{x,U}(z) \quad \text{for all } n \in \mathbb{N}_0.$$

The measure $\sigma_{x,U}$ is called a **spectral measure** of an element x . We will often denote $\sigma_{x,U}$ simply by σ_x .

Theorem 1.13 (Fourier Uniqueness Theorem, [Lemańczyk, thm. 1.14])

Let σ_1, σ_2 be two non-negative, finite Borel measures on \mathbb{T} . Then

$$\sigma_1 = \sigma_2 \iff \forall_{n \in \mathbb{Z}} \hat{\sigma}_1(n) = \hat{\sigma}_2(n).$$

Proposition 1.5

Let $U : H \rightarrow H$ be an isometry on Hilbert space H . For every $x \in H$ and finite sequence $(a_n)_{n=0}^N$ of complex numbers the following equality holds:

$$\left\| \sum_{n=0}^N a_n U^n x \right\|^2 = \int_{\mathbb{T}} \left| \sum_{n=0}^N a_n z^n \right|^2 d\sigma_x(z) = \left\| \sum_{n=0}^N a_n z^n \right\|_{L^2(\mathbb{T}, \mathcal{B}(\mathbb{T}), \sigma_x)}^2.$$

Proof: For any sequence $(r_n)_{n \in \mathbb{Z}}$ in Proposition 1.4, we have (by equalities (1.1) and (1.2))

$$\begin{aligned} \left\| \sum_{n=0}^N a_n U^n x \right\|^2 &= \sum_{n,m=0}^N r_{n-m} a_n \overline{a_m} = \sum_{n,m=0}^N a_n \overline{a_m} \int_{\mathbb{T}} z^{n-m} d\sigma_x(z) = \sum_{n,m=0}^N a_n \overline{a_m} \int_{\mathbb{T}} z^n \overline{z^m} d\sigma_x(z) = \\ &= \sum_{n=0}^N a_n \int_{\mathbb{T}} z^n \left(\sum_{m=0}^N \overline{a_m z^m} \right) d\sigma_x(z) = \int_{\mathbb{T}} \sum_{n=0}^N a_n z^n \left(\sum_{m=0}^N \overline{a_m z^m} \right) d\sigma_x(z) = \int_{\mathbb{T}} \left| \sum_{n=0}^N a_n z^n \right|^2 d\sigma_x(z). \end{aligned}$$

□

In order to prove Wiener's Criterion of Continuity, we need the following lemma (also due to Wiener):

Lemma 1.1 (Wiener, [Lemańczyk, lemma 1.16])

Let σ be a finite non-negative Borel measure on \mathbb{T} . Denote by $\{a_1, a_2, \dots\}$ the set of all atoms of the measure σ . Then

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\hat{\sigma}(n)|^2 = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\hat{\sigma}(-n)|^2 = \sum_{m \geq 1} \sigma(\{a_m\})^2.$$

Proof: Note first, that since $\widehat{\sigma}(n) = \overline{\widehat{\sigma}(-n)}$, thus the limits $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)|^2$ and $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(-n)|^2$ coincide (if they exists). Note further, that since the measure σ is finite, the series $\sum_{m \geq 1} \sigma(\{a_m\})^2$ converges (we know that $\sum_{m \geq 1} \sigma(\{a_m\}) < \infty$ and only for finitely many $m \in \mathbb{N}$ there can be $\sigma(\{a_m\}) \geq 1$). By Fubini's Theorem we have

$$\begin{aligned} |\widehat{\sigma}(n)|^2 &= \widehat{\sigma}(n) \overline{\widehat{\sigma}(n)} = \int_{\mathbb{T}} z^n d\sigma(z) \overline{\int_{\mathbb{T}} w^n d\sigma(w)} = \int_{\mathbb{T}} z^n \left(\int_{\mathbb{T}} \overline{w}^n d\sigma(w) \right) d\sigma(z) = \\ &= \int_{\mathbb{T} \times \mathbb{T}} (z\overline{w})^n d\sigma \otimes \sigma(z, w), \end{aligned}$$

and further

$$\frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)|^2 = \int_{\mathbb{T}^2} \frac{1}{N} \sum_{n=0}^{N-1} (z\overline{w})^n d\sigma \otimes \sigma(z, w). \quad (1.3)$$

For $z, w \in \mathbb{T}$ we have also $z\overline{w} \in \mathbb{T}$ and $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} (z\overline{w})^n = \mathbb{1}_{\{(z,w) \in \mathbb{T}^2 : z\overline{w}=1\}}(z, w) = \mathbb{1}_{\Delta}(z, w)$, where $\Delta = \{(z, w) \in \mathbb{T}^2 : z = w\}$. Since $|\frac{1}{N} \sum_{n=0}^{N-1} (z\overline{w})^n| \leq \frac{1}{N} \sum_{n=0}^{N-1} |(z\overline{w})^n| = 1$, we have by Lebesgue Dominated Convergence Theorem

$$\lim_{N \rightarrow \infty} \int_{\mathbb{T}^2} \frac{1}{N} \sum_{n=0}^{N-1} (z\overline{w})^n d\sigma \otimes \sigma(z, w) = \int_{\mathbb{T}^2} \mathbb{1}_{\Delta}(z, w) d\sigma \otimes \sigma(z, w). \quad (1.4)$$

By Fubini's Theorem we have

$$\begin{aligned} \int_{\mathbb{T}^2} \mathbb{1}_{\Delta}(z, w) d\sigma \otimes \sigma(z, w) &= \int_{\mathbb{T}} \left(\int_{\mathbb{T}} \mathbb{1}_{\Delta}(z, w) d\sigma(w) \right) d\sigma(z) = \int_{\mathbb{T}} \left(\int_{\mathbb{T}} \mathbb{1}_{\{z\}}(w) d\sigma(w) \right) d\sigma(z) = \\ &= \int_{\mathbb{T}} \sigma(\{z\}) d\sigma(z) = \int_{\bigcup_{m \geq 1} \{a_m\}} \sigma(\{z\}) d\sigma(z) = \sum_{m \geq 1} \sigma(\{a_m\})^2, \end{aligned}$$

what combined with (1.3) and (1.4) completes the proof. \square

Corollary 1.4 (Wiener's Criterion of Continuity)

Non-negative finite Borel measure σ on \mathbb{T} is continuous if and only if $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)|^2 =$

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(-n)|^2 = 0. \quad \square$$

Remark Recall the following inequality: for any $y_1, \dots, y_N \in \mathbb{R}$ we have

$$\left(\sum_{k=1}^N y_k \right)^2 \leq N \sum_{k=1}^N y_k^2. \quad (1.5)$$

It can be seen from the following:

$$\begin{aligned}
N \sum_{k=1}^N y_k^2 - \left(\sum_{k=1}^N y_k \right)^2 &= N \sum_{k=1}^N y_k^2 - \left(\sum_{k=1}^N y_k^2 + 2 \sum_{1 \leq i < j \leq N} y_i y_j \right) = \\
&= (N-1) \sum_{k=1}^N y_k^2 - 2 \sum_{1 \leq i < j \leq N} y_i y_j \geq \sum_{1 \leq i < j \leq N} (y_i - y_j)^2 \geq 0.
\end{aligned}$$

From (1.5) we can obtain another

Corollary 1.5

If non-negative finite Borel measure σ on \mathbb{T} is continuous, then $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)| = 0$.

Proof: By Corollary 1.4 we have $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)|^2 = 0$ and by (1.5) we have

$$\left(\frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)| \right)^2 \leq \frac{1}{N^2} \left(N \sum_{n=0}^{N-1} |\widehat{\sigma}(n)|^2 \right) = \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)|^2 \xrightarrow{N \rightarrow \infty} 0.$$

By the continuity of function $\mathbb{R} \ni x \mapsto x^2 \in \mathbb{R}$ we have also $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |\widehat{\sigma}(n)| = 0$. \square

After establishing von Neumann's Ergodic Theorem in next chapter, we will be able to prove another important lemma about spectral measures.

Chapter 2

Introduction to ergodic theory

This chapter includes short introduction to the ergodic theory. We give basic concepts and facts. The presentation is based on [Einsiedler, Ward] and [Eisner et al].

2.1. Measure preserving dynamical systems

The main object in ergodic theory is a measure preserving system. In the first part of the thesis we will consider only discrete time dynamical systems arising from single transformation.

Definition 2.1 (Measure preserving system)

Let (X, \mathcal{A}, μ) be a probability space. A measurable map $T : X \rightarrow X$ is called **measure preserving** (or **μ -invariant**) if

$$\forall_{A \in \mathcal{A}} \mu(T^{-1}A) = \mu(A).$$

In this case the measure μ is called **T -invariant** and (X, \mathcal{A}, μ, T) is called a **measure preserving (dynamical) system**.

Remark Sometimes it is enough to consider measurable dynamical system (X, \mathcal{A}, μ, T) without the assumption $\mu(T^{-1}A) = \mu(A)$ for $A \in \mathcal{A}$. Namely, we assume only measurability of T and that $\mu(A) \neq 0 \implies \mu(T^{-1}A) \neq 0$. However, in our thesis, we always assume that transformation is measure preserving. We assume also (for the sake of simplicity) that measure is already normalized ($\mu(X) = 1$), but the theory is valid also for finite measures. Some of the results can be established for not necessarily finite, but at least σ -finite measures.

Example 2.1 Consider dynamical system $(\mathbb{T}, \mathcal{B}(\mathbb{T}), m, R_\lambda)$, where \mathbb{T} is a unit circle on a complex plane, $\mathcal{B}(\mathbb{T})$ is a Borel σ -field on \mathbb{T} , m is a (normalized) Lebesgue measure on $\mathcal{B}(\mathbb{T})$ and $R_\lambda : \mathbb{T} \rightarrow \mathbb{T}$, $R_\lambda(z) = \lambda z$ is a rotation (multiplication) by $\lambda \in \mathbb{T}$. It is obvious that R_λ preserves (the Haar) measure of arcs, hence R_λ is measure preserving (clearly arcs generate $\mathcal{B}(\mathbb{T})$ and it's easy to notice that it is enough to check preservation of measure on the generator of given σ -field). This is an important example, because multiplication by complex number from a unit circle occur in the Wiener-Wintner type theorems.

Fact 2.1 ([Einsiedler, Ward, lemma 2.6])

Let (X, \mathcal{A}, μ) be a probability space. A measurable map $T : X \rightarrow X$ is measure preserving if and

only if for every $f \in \mathcal{L}^\infty(\mu)$ we have

$$\int_X f(x) d\mu(x) = \int_X f(Tx) d\mu(x). \quad (2.1)$$

Moreover, if T is measure preserving, then (2.1) holds for all $f \in \mathcal{L}^1(\mu)$.

For two measure preserving systems it is natural to consider their product:

Fact 2.2 Let (X, \mathcal{A}, μ, T) and (Y, \mathcal{C}, ν, S) be measure preserving systems. Then the system $(X \times Y, \mathcal{A} \otimes \mathcal{C}, \mu \otimes \nu, T \times S)$ with $T \times S(x, y) := (Tx, Sy)$ is also a measure preserving system which is called the **direct product** of (X, \mathcal{A}, μ, T) and (Y, \mathcal{C}, ν, S) .

We will give now definition of the Koopman Operator, which gives possibility to apply the functional analysis in ergodic theory.

Definition 2.2 (Koopman Operator on $L^p(\mu)$)

Let (X, \mathcal{A}, μ, T) be a measure preserving system. For $1 \leq p < \infty$ we define the **Koopman Operator on $L^p(\mu)$** (induced by T) as $U_T : L^p(\mu) \rightarrow L^p(\mu)$ is defined by

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

Remark Note that since $f \in L^p(\mu)$ is formally an equivalence class of functions (equal almost everywhere), it doesn't make sense to consider the superposition $f \circ T(x) := f(Tx)$ for a fixed point $x \in X$. On the other hand, note that if for $f, g \in \mathcal{L}^p(\mu)$ we have $f = g$ almost everywhere, then also $f \circ T = g \circ T$ almost everywhere. Indeed, $\mu(\{f \circ T \neq g \circ T\}) = \mu(\{x \in X : f(Tx) \neq g(Tx)\}) = \mu(T^{-1}(\{f \neq g\})) = \mu(\{f \neq g\}) = 0$, since T is measure preserving. This shows that the equivalence class of $f \circ T$ is uniquely determined by the equivalence class of f , so it makes sense to define $f \circ T$ for $f \in L^p(\mu)$. Note further, that for $f \in L^p(\mu)$ we have by 2.1

$$\int_X |f \circ T|^p d\mu = \int_X |f|^p \circ T d\mu = \int_X |f|^p d\mu < \infty,$$

so Koopman Operator is well defined.

Fact 2.3 For a measure preserving system (X, \mathcal{A}, μ, T) , its Koopman Operator $U_T : L^p(\mu) \rightarrow L^p(\mu)$, $1 \leq p < \infty$ is an isometry. In particular, for $p = 2$ we have $\langle U_T f, U_T g \rangle = \langle f, g \rangle$ for $f, g \in L^2(\mu)$.

The most important class of measure preserving systems are ergodic dynamical systems.

Definition 2.3 A measure preserving system (X, \mathcal{A}, μ, T) is called an **ergodic (dynamical system)** if

$$\forall_{A \in \mathcal{A}} [T^{-1}A = A \implies \mu(A) \in \{0, 1\}].$$

In the above situation, the transformation T and the measure μ are also called **ergodic**.

Set $A \in \mathcal{A}$ with $T^{-1}A = A$ is called a **T -invariant set** (or simply invariant set). Thus, the ergodicity of the system means that only null sets (sets of zero measure) and full measure sets are invariant.

Example 2.2 Consider the rotation system $(\mathbb{T}, \mathcal{B}(\mathbb{T}), m, R_\lambda)$ from Example 2.1. It may be proved that it is ergodic if and only if $\text{Arg}(\lambda) \notin 2\pi\mathbb{Q}$, where $\text{Arg}(z)$ stands for the argument of complex number z .

We now give a useful characterization of ergodicity. Namely,

Proposition 2.1

A measure preserving system (X, \mathcal{A}, μ, T) is ergodic if and only if for some (every) $p \in [1, \infty)$ we have

$$\forall_{f \in L^p(\mu)} [f \circ T = f \text{ } \mu\text{-a.e.} \implies f \text{ is equal to a constant function } \mu\text{-a.e.}].$$

Using the above characterization, we will give some spectral properties of Koopman Operator on $L^2(\mu)$.

Proposition 2.2

Let (X, \mathcal{A}, μ, T) be a measure preserving system and $U_T : L^2(\mu) \rightarrow L^2(\mu)$ its Koopman Operator on $L^2(\mu)$. Then

- (1) $\sigma(U_T) \subset \mathbb{T}$,
- (2) If T is ergodic then for every eigenfunction $f \in L^2(\mu)$ of U_T we have $|f| = \text{const } \mu\text{-a.e.}$,
- (3) If T is ergodic then for every eigenvalue $\lambda \in \sigma(U_T)$ its eigenspace is one-dimensional.

Proof: (1) Suppose that for $f \in L^2(\mu)$, $f \neq 0$, $\lambda \in \mathbb{C}$ we have $U_T f = \lambda f$. Since U_T is an isometry we have $\|f\|_2 = \|U_T f\|_2 = \|\lambda f\|_2 = |\lambda| \|f\|_2$. Since $f \neq 0 \implies \|f\|_2 \neq 0$, we get $|\lambda| = 1$, so $\lambda \in \mathbb{T}$.

(2) Suppose that $U_T f = \lambda f$. By (1) we have $|\lambda| = 1$, so $|f| \circ T = |f|$. Since $U_T f = \lambda f$, we have $|f| \circ T = |f|$, so $|f|$ is T -invariant. T is ergodic, hence by Proposition 2.1 $|f| = \text{const } \mu\text{-a.e.}$

(3) Take $f, g \in H_\lambda$ and assume that $f \neq 0$. By (2) we have $|f| = \text{const } \mu\text{-a.e.}$, hence there must be also $|f| \neq 0 \text{ } \mu\text{-a.e.}$ and further $f \neq 0 \text{ } \mu\text{-a.e.}$. Since also $|g| = \text{const } \mu\text{-a.e.}$, we have $\frac{|g|}{|f|} = \text{const } \mu\text{-a.e.}$, so $\frac{g}{f} \in L^2(\mu)$. Now we have $U_T(\frac{g}{f}) = \frac{g}{f} \circ T = \frac{g \circ T}{f \circ T} = \frac{\lambda g}{\lambda f} = \frac{g}{f}$. By ergodicity, there exists $\alpha \in \mathbb{C}$ such that $\frac{g}{f} = \alpha \text{ } \mu\text{-a.e.}$, so $g = \alpha f$, hence H_λ is one-dimensional. \square

One of the main interests of the ergodic theory is the asymptotic behavior of ergodic averages $\frac{1}{N} \sum_{n=0}^{N-1} f(T^n x)$. The main and classical result in this field is the celebrated Birkhoff's Ergodic Theorem.

Theorem 2.1 (Birkhoff's Ergodic Theorem [Einsiedler, Ward, thm. 2.30])

Let (X, \mathcal{A}, μ, T) be a measure preserving system. If $f \in \mathcal{L}^1(\mu)$, then

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} f(T^n x) = f^*(x), \text{ } \mu\text{-a.e. and in } L^1(\mu),$$

where $f^* \in \mathcal{L}^1(\mu)$ is a T -invariant function with

$$\int_X f^* d\mu = \int_X f d\mu.$$

If T is ergodic, then

$$f^*(x) = \int_X f d\mu \quad \mu\text{-a.e.}$$

Remark Birkhoff's Ergodic Theorem is often stated for $f \in L^1(\mu)$ instead of $f \in \mathcal{L}^1(\mu)$, although it requires evaluation of the function on the orbit of a point $x \in X$. In this situation we understand it as follows: for every function in $\mathcal{L}^1(\mu)$ from the equivalence class $f \in L^1(\mu)$ there is a almost sure convergence. Some of the further pointwise ergodic theorems will be also stated in this fashion.

2.2. Topological dynamical systems

It is possible to consider measurable dynamical systems with some additional structure on the phase space X . In this section we will give brief introduction to topological systems, in which the phase space will be a compact topological space. We assume compact spaces to be Hausdorff.

Definition 2.4 Pair (X, T) consisting of a compact topological space X and continuous map $T : X \rightarrow X$ is called a **topological dynamical system**.

Example 2.3 Note that system (\mathbb{T}, R_λ) for $\lambda \in \mathbb{T}$ is a topological dynamical system.

For a topological dynamical system (X, T) we will consider set $\mathcal{M}(X)$ of finite, regular, complex-valued measures on X as set of natural measures on X . When considering topological dynamical systems, we will assume all measures to be regular and Borel. Denote by $\mathcal{M}^1(X)$ set of all non-negative, probability measures from $\mathcal{M}(X)$. We will also consider set of all T -invariant measures from $\mathcal{M}^1(X)$, denoted by $\mathcal{M}^T(X)$. By $\mathcal{E}^T(X)$ we will denote set of all ergodic measures from $\mathcal{M}^1(X)$. The following theorem makes use of the Banach-Alaoglu Theorem and Krein-Milman Theorem to give some properties of these sets.

Theorem 2.2 ([Eisner et al])

Let (X, T) be a topological dynamical system. Then

- (1) $\mathcal{E}^T(X) \subset \mathcal{M}^T(X) \subset \mathcal{M}^1(X) \subset \overline{B}(0, 1)$, where $\overline{B}(0, 1)$ is a unit ball in a space $\mathcal{M}(X)$ with the variation norm,
- (2) the sets $\mathcal{M}^T(X)$ and $\mathcal{M}^1(X)$ are convex, weak* compact and if X is metrizable, then weak* sequentially compact,
- (3) $\mathcal{E}^T(X) = \text{Ext}(\mathcal{M}^T(X))$,
- (4) $\mathcal{M}^T(X) = \overline{\text{co}}(\mathcal{E}^T(X))$.

For a given topological dynamical system (X, T) define a map $T_* : \mathcal{M}^1(X) \rightarrow \mathcal{M}^1(X)$ by $T_*\mu(A) := \mu(T^{-1}A)$ for $A \in \mathcal{B}(X)$. Note that $T_*\mu$ is simply the transport of measure μ by the map T . We have

$$\int_X f \circ T d\mu = \int_X f dT_*\mu.$$

Lemma 2.1 ([Einsiedler, Ward, thm. 4.1])

Let (X, T) be a topological dynamical system with metrizable X and let $(\nu)_{n \in \mathbb{N}}$ be a sequence of measures from $\mathcal{M}^1(X)$. Then any weak* cluster point μ of the sequence $\mu_N := \frac{1}{N} \sum_{n=0}^N T_*^n \nu_N$, $N \in \mathbb{N}$ is a T -invariant measure.

Proof: Suppose that there is a subsequence $(\mu_{N_j})_{j \in \mathbb{N}}$ with $\mu_{N_j} \xrightarrow{*w} \mu$. It is enough to show $\int_X f dT_*\mu = \int_X f d\mu$ (for every $f \in C(X)$), it will mean that μ and $T_*\mu$ give rise to the same functional on $C(X)$, hence must be equal). We have

$$\begin{aligned} & \left| \int_X f d\mu - \int_X f dT_*\mu \right| = \left| \int_X f d\mu - \int_X f \circ T d\mu \right| = \\ & = \left| \lim_{j \rightarrow \infty} \int_X f d \frac{1}{N_j} \sum_{n=0}^{N_j-1} T_*^n \nu_{N_j} - \lim_{j \rightarrow \infty} \int_X f \circ T d \frac{1}{N_j} \sum_{n=0}^{N_j-1} T_*^n \nu_{N_j} \right| = \\ & = \lim_{j \rightarrow \infty} \frac{1}{N_j} \left| \sum_{n=0}^{N_j-1} \int_X f \circ T^n d\nu_{N_j} - \sum_{n=0}^{N_j-1} \int_X f \circ T^{n+1} d\nu_{N_j} \right| = \\ & = \lim_{j \rightarrow \infty} \frac{1}{N_j} \left| \int_X f d\nu_{N_j} - \int_X f \circ T^{N_j} d\nu_{N_j} \right| \leq \lim_{j \rightarrow \infty} \frac{2\|f\|_{\sup}}{N_j} = 0. \end{aligned}$$

Note that since $\mathcal{M}^1(X)$ weak* sequentially compact, thus the sequence $(\mu_N)_{N \in \mathbb{N}}$ from the above lemma always has a weak* cluster point. This proves the following remarkable theorem.

Theorem 2.3 (Krylov–Bogoljubov[Eisner et al, thm. 10.2])

For every topological dynamical system (X, T) we have $\mathcal{M}^T(X) \neq \emptyset$, i.e. there exists at least one T -invariant, regular, probability Borel measure μ on X .

Since $\mathcal{M}^T(X) = \overline{\text{co}}(\mathcal{E}^T(X))$, there exists an ergodic probability measure for (X, T) and there is exactly one invariant probability if and only if there is exactly one ergodic probability. The unique invariant measure is automatically ergodic. This leads to the following

Definition 2.5 We say that topological dynamical system (X, T) is **uniquely ergodic** if it admits exactly one invariant probability measure, i.e. $|\mathcal{M}^T(X)| = 1$.

It follows (apply the Birkhoff ergodic theorem) that

Theorem 2.4 ([Eisner et al, thm. 10.6])

For a topological dynamical system (X, T) the following conditions are equivalent:

- (1) (X, T) is uniquely ergodic,
- (2) for every $f \in C(X)$ there is a constant $c(f) \in \mathbb{C}$ such that

$$\forall_{x \in X} \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} f(T^n x) = c(f),$$

(3) for every $f \in C(X)$ there is a constant $c(f) \in \mathbb{C}$ such that

$$\frac{1}{N} \sum_{n=0}^{N-1} f \circ T^n \xrightarrow{\|\cdot\|_{\text{sup}}} c(f).$$

Under any of the above assumptions, $c(f) = \int_X f d\mu$, where μ is the unique ergodic probability measure.

Definition 2.6 Let X be a compact topological space and take a non-negative measure $\mu \in \mathcal{M}(X)$. We define the **support of a measure** μ as $\text{supp}(\mu) = \{x \in X : \text{for every open set } U \text{ with } x \in U \text{ there is } \mu(U) > 0\}$.

Fact 2.4 Let (X, T) be a topological dynamical system. For every $\mu \in \mathcal{M}^1(X)$

- (1) $\text{supp}(\mu)$ is closed,
- (2) $\mu(\text{supp}(\mu)) = 1$,
- (3) if μ is invariant, then $T(\text{supp}(\mu)) \subset \text{supp}(\mu)$,
- (4) for every $f \in L^1(\mu)$ there is

$$\int_X f d\mu = \int_{\text{supp}(\mu)} f d\mu,$$

- (5) for $f, g \in C(X)$ there is $f = g$ μ -a.e. if and only if $f(x) = g(x)$ for every $x \in \text{supp}(\mu)$,
- (6) if $F \subset X$ is closed and $\mu(F) = 1$ then $\text{supp}(\mu) \subset F$,
- (7) if $(\mu_\alpha)_{\alpha \in \Lambda}$ is a net of measures from $\mathcal{M}^1(X)$ with $\text{supp}(\mu_\alpha) \subset F$ for some closed F and every $\alpha \in \Lambda$, then $\mu_\alpha \xrightarrow{w^*} \mu \in \mathcal{M}^1(X)$ implies that $\text{supp}(\mu) \subset F$.

Proof: (1) If $x \in X \setminus \text{supp}(\mu)$ then there exists an open set U with $x \in U$ such that $\mu(U) = 0$. Hence $U \subset X \setminus \text{supp}(\mu)$, what proves that $X \setminus \text{supp}(\mu)$ is open. It follows that $\text{supp}(\mu)$ is closed.

(2) It's enough to show that $X \setminus \text{supp}(\mu)$ has zero measure. By regularity, we know that

$$\mu(X \setminus \text{supp}(\mu)) = \sup\{\mu(K) : K \subset X \setminus \text{supp}(\mu), K \text{ is compact}\}. \quad (2.2)$$

Take any compact set K with $K \subset X \setminus \text{supp}(\mu)$. For every $x \in K$ there exists open U_x with $x \in U_x$ and $\mu(U_x) = 0$ (by (1)). Now we have $K \subset \bigcup_{x \in K} U_x$ and since K is compact, we can

choose x_1, \dots, x_n such that $K \subset \bigcup_{j=1}^n U_{x_j}$, so $\mu(K) \leq \sum_{j=1}^n \mu(U_{x_j}) = 0$, so $\mu(K) = 0$. By (2.2)

we have $\mu(X \setminus \text{supp}(\mu)) = 0$.

(3) We have to show that if $x \in \text{supp}(\mu)$ then $Tx \in \text{supp}(\mu)$. Assume that $Tx \notin \text{supp}(\mu)$. Then there exists an open set U with $Tx \in U$ and $\mu(U) = 0$. Now we have $x \in T^{-1}U$, $T^{-1}U$ is open (since T is continuous) and $\mu(T^{-1}U) = \mu(U) = 0$. In particular $x \notin \text{supp}(\mu)$, a contradiction.

(4) Follows from (2).

(5) If $f = g$ on $\text{supp}(\mu)$ then (2) implies that $f = g$ μ -a.e. On the other hand, if there is $x \in \text{supp}(\mu)$ with $f(x) \neq g(x)$, then there must also be $f \neq g$ on some open neighborhood U of x . Since $x \in \text{supp}(\mu)$, there is $\mu(U) > 0$, hence $\mu(\{f \neq g\}) \geq \mu(U) > 0$.

(6) Suppose that there exists $x \in \text{supp}(\mu)$ with $x \in F'$. Since F' is open there is $\mu(F') > 0$, what contradicts $\mu(F) = 1$.

(7) By the previous point, it suffices to show that $\mu(F) = 1$. Let us fix $\varepsilon > 0$. By the regularity of μ there exists open set U with $F \subset U$ and $\mu(U) < \mu(F) + \varepsilon$. By the Urysohn's lemma (Theorem 1.11) there exists continuous function $f : X \rightarrow [0, 1]$ with $f = 1$ on F and $f = 0$ on U' . We have

$$1 = \mu_\alpha(F) \leq \int_X f d\mu_\alpha \rightarrow \int_X f d\mu = \int_U f d\mu \leq \mu(U) \|f\|_{\text{sup}} = \mu(U) < \mu(F) + \varepsilon.$$

Passing with ε to zero finishes the proof. \square

Since T is continuous, we can consider the Koopman Operator on space $C(X)$, namely $U_T : C(X) \rightarrow C(X)$ with $U_T f = f \circ T$. U_T is a positive operator, it's a contraction and $U_T(\mathbf{1}) = \mathbf{1}$.

2.3. von Neumann's Ergodic Theorem

In this section we state von Neumann's (Mean) Ergodic Theorem, which can be seen as a first operator theoretic type ergodic theorem. It follows from the more general Corollary 4.3. We also prove its consequences in spectral theory for isometries.

Theorem 2.5 (von Neumann's Ergodic Theorem [Weber, thm. 1.3.1])

Let $U : H \rightarrow H$ be a contraction on a complex Hilbert space H . Then for every $f \in H$ there is

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} U^n f = Pf \text{ (in the norm } \|\cdot\|),$$

where $P : H \rightarrow H$ is an orthogonal projection on the closed subspace of U -invariant vectors $H_U = \{g \in H : Ug = g\}$. Moreover,

$$H = H_U \oplus H_0,$$

where $H_0 = \overline{\{g - Ug : g \in H\}}$.

Corollary 2.1

Let $U : H \rightarrow H$ be an isometry on a Hilbert space H and take $f \in H, \lambda \in \mathbb{T}$. Then $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n U^n f = P_\lambda f$, where P_λ is an orthogonal projection to the H_λ - the eigenspace of $\lambda \in \mathbb{T}$.

Proof: Note that the operator $V : H \rightarrow H$ defined by $V := \lambda U$ is also an isometry, since $\langle Vf, Vg \rangle = \langle \lambda Uf, \lambda Ug \rangle = \lambda \bar{\lambda} \langle Uf, Ug \rangle = |\lambda|^2 \langle f, g \rangle = \langle f, g \rangle$. By von Neumann's Theorem we have that

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n U^n f = \frac{1}{N} \sum_{n=0}^{N-1} V^n f \longrightarrow Qf,$$

where Q is an orthogonal projection on a subspace $\{f \in H : Vf = f\} = \{f \in H : \lambda Uf = f\} = \{f \in H : Uf = \bar{\lambda}f\} = H_{\bar{\lambda}}$. It follows that $Q = P_{\bar{\lambda}}$. \square

Lemma 2.2

Let $U : H \rightarrow H$ be an isometry on Hilbert space H and $f \in H$. Then $\sigma_f(\{\lambda\}) = \|P_{\lambda}f\|^2$, where σ_f denotes the spectral measure of f .

Proof: From Corollary 2.1 we have

$$\left\| \frac{1}{N} \sum_{n=0}^{N-1} \bar{\lambda}^n U^n f \right\|^2 \rightarrow \|P_{\lambda}f\|^2, \quad (2.3)$$

but from Proposition 1.5 we have also

$$\left\| \frac{1}{N} \sum_{n=0}^{N-1} \bar{\lambda}^n U^n f \right\|^2 = \int_{\mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \bar{\lambda}^n z^n \right|^2 d\sigma_f(z) = \int_{\mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \left(\frac{z}{\lambda} \right)^n \right|^2 d\sigma_f(z). \quad (2.4)$$

Note that for every $z \in \mathbb{T}$ we have $\frac{1}{N} \sum_{n=0}^{N-1} \left(\frac{z}{\lambda} \right)^n \rightarrow \mathbb{1}_{\{\lambda\}}(z)$, hence $\left| \frac{1}{N} \sum_{n=0}^{N-1} \left(\frac{z}{\lambda} \right)^n \right|^2 \rightarrow |\mathbb{1}_{\{\lambda\}}(z)|^2 = \mathbb{1}_{\{\lambda\}}(z)$. Since $\left| \frac{1}{N} \sum_{n=0}^{N-1} \left(\frac{z}{\lambda} \right)^n \right|^2 \leq \left(\frac{1}{N} \sum_{n=0}^{N-1} \left| \frac{z}{\lambda} \right|^n \right)^2 = 1$, we can make use of Lebesgue's Dominated Convergence Theorem and obtain

$$\int_{\mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \left(\frac{z}{\lambda} \right)^n \right|^2 d\sigma_f(z) \longrightarrow \int_{\mathbb{T}} \mathbb{1}_{\{\lambda\}}(z) d\sigma_f(z) = \sigma_f(\{\lambda\}). \quad (2.5)$$

Putting together (2.3), (2.4) and (2.5) the proof is completed. \square

Chapter 3

Wiener-Wintner theorems for deterministic transformations

In this chapter we introduce and prove pointwise Wiener-Wintner type theorems. We start with stating classical Wiener-Wintner theorem, which is a modification of Birkhoff's Ergodic Theorem. It was originally stated by Wiener and Wintner in 1941 ([WW]).

Theorem 3.1 (Wiener-Wintner ergodic theorem, [Assani, thm. 2.3])

Let (X, \mathcal{A}, μ, T) be an ergodic dynamical system and fix function $f \in \mathcal{L}^1(\mu)$. There exists a measurable set X_f of full measure ($\mu(X_f) = 1$) such that for each $x \in X_f$ the averages

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \quad (3.1)$$

converge for all $\lambda \in \mathbb{T}$.

It will be useful for us to use the following

Definition 3.1 (Wiener-Wintner property, [Assani, def. 2.7])

Let (X, \mathcal{A}, μ, T) be a measurable dynamical system. A function $f \in L^1(\mu)$ is said to satisfy the Wiener-Wintner property if there exists a set X_f of full measure such that for each $x \in X_f$ the limit

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

exists for all $\lambda \in \mathbb{T}$.

Using the notion of Wiener-Wintner property, the Theorem 3.1 can be restated as follows: if (X, \mathcal{A}, μ, T) is an ergodic dynamical system, then every $f \in L^1(\mu)$ has a Wiener-Wintner property.

Remark Note that for a fixed $\lambda \in \mathbb{T}$ it is easy to achieve a.e. convergence in (3.1). Take a product system $(X \times \mathbb{T}, \mathcal{A} \otimes \mathcal{B}(\mathbb{T}), \mu \otimes m, T \times R_\lambda)$ and observe that it is measure preserving since both (X, \mathcal{A}, μ, T) and $(\mathbb{T}, \mathcal{B}(\mathbb{T}), m, R_\lambda)$ are measure preserving. Define a function $g : X \times \mathbb{T} \rightarrow \mathbb{C}$ by $g(x, \omega) = \omega f(x)$. We have $g \in \mathcal{L}^1(\mu \otimes m)$ since, by Fubini's Theorem,

$$\int_{X \times \mathbb{T}} |g(x, \omega)| d\mu \otimes m(x, \omega) = \int_{X \times \mathbb{T}} |\omega| |f(x)| d\mu \otimes m(x, \omega) = \int_{X \times \mathbb{T}} |f(x)| d\mu \otimes m(x, \omega) =$$

$$= \int_X |f(x)| d\mu(x) < \infty.$$

By Birkhoff's Ergodic Theorem the averages

$$\frac{1}{N} \sum_{n=0}^{N-1} g(T^n x, R_\lambda^n \omega) = \frac{1}{N} \sum_{n=0}^{N-1} g(T^n x, \lambda^n \omega) = \frac{1}{N} \sum_{n=0}^{N-1} \omega \lambda^n f(T^n x)$$

converge for $\mu \otimes m$ almost all pairs $(x, \omega) \in X \times \mathbb{T}$ and (since $\omega \neq 0$) also

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x)$$

converge $\mu \otimes m$ a.e. The last limit does not depend on ω , so this implies μ a.e. convergence of sequence (3.1). Further, for a countable subset $C \subset \mathbb{T}$, we can find a set X_f such that the sequence in (3.1) is still convergent for all $x \in X_f$ and $\lambda \in C$ (it is enough to take for X_f an intersection of countably many sets of full measure on which we have convergence for fixed $\lambda \in C$). This shows that the difficulty in Wiener-Wintner theorem is obtaining a set of full measure on which convergence will hold for all (uncountably many) $\lambda \in \mathbb{T}$.

Three proofs of this theorem can be found in [Assani]. We present one of them, which main ingredient is itself a generalization of Wiener-Wintner theorem - its uniform version due to J. Bourgain. Our proofs are taken from [Assani], although they are slightly modified in a way which doesn't require the assumption of separability of the space $L^2(\mu)$.

3.1. Bourgain's uniform Wiener-Wintner theorem

In order to state the theorem, we need to introduce the notion of Kronecker factor.

Definition 3.2 (Kronecker factor, [Assani, def. 2.5])

Let (X, \mathcal{A}, μ, T) be a measure preserving system and let $U_T : L^2(\mu) \rightarrow L^2(\mu)$ be its Koopman operator on $L^2(\mu)$. **Kronecker factor** $\mathcal{K} \subset L^2(\mu)$ is a closure (in $L^2(\mu)$) of a linear subspace spanned by eigenfunctions of U_T , i.e.

$$\mathcal{K} := \overline{\text{span}} \{ f \in L^2(\mu) : f \circ T = \lambda f \text{ for some } \lambda \in \mathbb{C} \}.$$

The closure is taken in $L^2(\mu)$ norm.

Theorem 3.2 (Bourgain's uniform Wiener-Wintner theorem [Assani, thm. 2.4])

Let (X, \mathcal{A}, μ, T) be an ergodic dynamical system and $f \in \mathcal{K}^\perp$. Then for μ a.e. $x \in X$ we have

$$\lim_{N \rightarrow \infty} \sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right| = 0.$$

For the proof we will need two following lemmas:

Lemma 3.1 ([Assani, prop. 2.2])

Let (X, \mathcal{A}, μ, T) be a measure preserving dynamical system. A function $f \in L^2(\mu)$ belongs to \mathcal{K}^\perp if and only if its spectral measure $\sigma_f := \sigma_{f, U_T}$ is continuous.

Proof: Let us fix $f \in \mathcal{K}^\perp$. Since for every $\lambda \in \mathbb{T}$ we have $H_\lambda \subset \mathcal{K}$ and f is orthogonal to \mathcal{K} , f must be also orthogonal to H_λ . If P_λ is an orthogonal projection to H_λ , then we have $P_\lambda f = 0$. By Lemma 2.2 we have $\sigma_f(\{\lambda\}) = \|P_\lambda f\|^2$ for all $\lambda \in \mathbb{T}$, so $\sigma_f(\{\lambda\}) = 0$ for all $\lambda \in \mathbb{T}$ and the measure σ_f is continuous. Conversely, fix $f \in L^2(\mu)$ and assume that σ_f is continuous. Then again by Lemma 2.2 we have $\|P_\lambda f\| = 0$, hence $f \in H_\lambda^\perp$ for every $\lambda \in \mathbb{T}$. So f is orthogonal to every eigenfunction of the operator U_T . We have (by linearity of the inner product) that f is orthogonal also to $\text{span}\{f \in L^2(\mu) : f \circ T = \lambda f \text{ for some } \lambda \in \mathbb{C}\}$ and finally (by continuity of the inner product) $f \in \mathcal{K}^\perp$. \square

Lemma 3.2 (Van der Corput inequality, [Weber, thm. 1.7.1])

Let H be a complex Hilbert space. For every finite sequence $x_0, x_1, \dots, x_{N-1} \in H$ and integer $R \in \{0, 1, \dots, N-1\}$ the following inequality holds:

$$\left\| \frac{1}{N} \sum_{n=0}^{N-1} x_n \right\|^2 \leq \leq \frac{N+R}{N(R+1)} \left(\frac{1}{N} \sum_{n=0}^{N-1} \|x_n\|^2 + \frac{1}{N(R+1)} \sum_{c=1}^R (R-c+1) \sum_{j=0}^{N-c-1} (\langle x_{j+c}, x_j \rangle + \langle x_j, x_{j+c} \rangle) \right).$$

If $H = \mathbb{C}$, this inequality becomes

$$\left| \frac{1}{N} \sum_{n=0}^{N-1} x_n \right|^2 \leq \frac{N+R}{N(R+1)} \left(\frac{1}{N} \sum_{n=0}^{N-1} |x_n|^2 + \frac{2}{N(R+1)} \sum_{c=1}^R (R-c+1) \text{Re} \left(\sum_{j=0}^{N-c-1} x_j \overline{x_{j+c}} \right) \right).$$

Proof: Let's make a convention that $x_n := 0$ for $n < 0$ and $n \geq N$. Observe that

$$\begin{aligned} \sum_{k=-R}^{N-1} \sum_{r=0}^R x_{k+r} &= (x_0) + (x_0 + x_1) + (x_0 + x_1 + x_2) + \dots + (x_0 + x_1 + \dots + x_R) + \\ &\quad + (x_1 + x_2 + \dots + x_{R+1}) + \dots + (x_{N-R-1} + x_{N-R} + \dots + x_{N-1}) + \\ &\quad + (x_{N-R} + x_{N-R+1} + \dots + x_{N-1}) + \dots + (x_{N-2} + x_{N-1}) + (x_{N-1}) = (R+1) \sum_{n=0}^{N-1} x_n. \end{aligned} \quad (3.2)$$

Using (3.2) together with inequality (1.5) for $y_k = \left\| \frac{1}{R+1} \sum_{r=0}^R x_{k+r} \right\|$, $-R \leq k \leq N-1$ we obtain

$$\begin{aligned} \left\| \sum_{n=0}^{N-1} x_n \right\| &= \left\| \sum_{k=-R}^{N-1} \frac{1}{R+1} \sum_{r=0}^R x_{k+r} \right\| \leq \sum_{k=-R}^{N-1} \left\| \frac{1}{R+1} \sum_{r=0}^R x_{k+r} \right\| \leq \\ &\leq (N+R)^{\frac{1}{2}} \left(\sum_{k=-R}^{N-1} \left\| \frac{1}{R+1} \sum_{r=0}^R x_{k+r} \right\|^2 \right)^{\frac{1}{2}} \end{aligned}$$

and further

$$\left\| \frac{1}{N} \sum_{n=0}^{N-1} x_n \right\|^2 \leq \frac{N+R}{N^2} \left(\sum_{k=-R}^{N-1} \left\| \frac{1}{R+1} \sum_{r=0}^R x_{k+r} \right\|^2 \right) = \frac{N+R}{N^2(R+1)^2} \left(\sum_{k=-R}^{N-1} \left\| \sum_{r=0}^R x_{k+r} \right\|^2 \right). \quad (3.3)$$

Let's write $[x, y] := \langle x, y \rangle + \langle y, x \rangle$. Now we have (using argument from (3.2))

$$\begin{aligned}
\sum_{k=-R}^{N-1} \left\| \sum_{r=0}^R x_{k+r} \right\|^2 &= \sum_{k=-R}^{N-1} \left\langle \sum_{r=0}^R x_{k+r}, \sum_{r=0}^R x_{k+r} \right\rangle = \sum_{k=-R}^{N-1} \sum_{s=0}^R \sum_{r=0}^R \langle x_{k+s}, x_{k+r} \rangle = \\
&= \sum_{k=-R}^{N-1} \left(\sum_{r=0}^R \|x_{k+r}\|^2 + \sum_{0 \leq s < r \leq R} (\langle x_{k+s}, x_{k+r} \rangle + \langle x_{k+r}, x_{k+s} \rangle) \right) = \\
&= (R+1) \sum_{n=0}^{N-1} \|x_n\|^2 + \sum_{k=-R}^{N-1} \sum_{0 \leq s < r \leq R} [x_{k+r}, x_{k+s}].
\end{aligned} \tag{3.4}$$

Since we've made a convention that $x_n = 0$ for $n < 0$ and $n \geq N$, we have that $[x_{k+r}, x_{k+s}] = 0$ for $k+s < 0$ or $k+s > N-1$ or $k+r < 0$ or $k+r > N-1$. It implies that it's enough to take the last summation in (3.4) over triples k, s, r with $s < r$ such that $0 \leq k+s \leq N-1 \wedge 0 \leq k+r \leq N-1$, which is equivalent to $-s \leq k \leq N-s-1 \wedge -r \leq k \leq N-r-1$ which is again (since $s < r$) equivalent to $-s \leq k \leq N-r-1$, so we have

$$\begin{aligned}
\sum_{k=-R}^{N-1} \sum_{0 \leq s < r \leq R} [x_{k+r}, x_{k+s}] &= \sum_{0 \leq s < r \leq R} \sum_{k=-R}^{N-1} [x_{k+r}, x_{k+s}] = \sum_{0 \leq s < r \leq R} \sum_{k=-s}^{N-r-1} [x_{k+r}, x_{k+s}] = \\
&\stackrel{j:=k+s}{=} \sum_{0 \leq s < r \leq R} \sum_{j=0}^{N-(r-s)-1} [x_{j+(r-s)}, x_j].
\end{aligned}$$

Note that the inner sum depends now only on the difference $r-s$, so by noting that $r-s = c$ for exactly $(R-c+1)$ pairs r, s such that $0 \leq s < r \leq R$ (where $1 \leq c \leq R$) we may continue to obtain

$$\sum_{k=-R}^{N-1} \sum_{0 \leq s < r \leq R} [x_{k+r}, x_{k+s}] \stackrel{c:=r-s}{=} \sum_{c=1}^R (R-c+1) \sum_{j=0}^{N-c-1} [x_{j+c}, x_j]. \tag{3.5}$$

Combining together (3.3), (3.4) and (3.5) we get to the conclusion

$$\begin{aligned}
\left\| \frac{1}{N} \sum_{n=0}^{N-1} x_n \right\|^2 &\leq \frac{N+R}{N^2(R+1)^2} \left(\sum_{k=-R}^{N-1} \left\| \sum_{r=0}^R x_{k+r} \right\|^2 \right) \\
&= \frac{N+R}{N^2(R+1)^2} \left((R+1) \sum_{n=0}^{N-1} \|x_n\|^2 + \sum_{k=-R}^{N-1} \sum_{0 \leq s < r \leq R} [x_{k+r}, x_{k+s}] \right) = \\
&= \frac{N+R}{N^2(R+1)^2} \left((R+1) \sum_{n=0}^{N-1} \|x_n\|^2 + \sum_{c=1}^R (R-c+1) \sum_{j=0}^{N-c-1} [x_{j+c}, x_j] \right) = \\
&= \frac{N+R}{N(R+1)} \left(\frac{1}{N} \sum_{n=0}^{N-1} \|x_n\|^2 + \frac{1}{N(R+1)} \sum_{c=1}^R (R-c+1) \sum_{j=0}^{N-c-1} (\langle x_{j+c}, x_j \rangle + \langle x_j, x_{j+c} \rangle) \right).
\end{aligned}$$

Inequality for $H = \mathbb{C}$ is immediate by observing that

$$\langle x_{j+c}, x_j \rangle + \langle x_j, x_{j+c} \rangle = x_{j+c} \overline{x_j} + x_j \overline{x_{j+c}} = 2\operatorname{Re}(x_j \overline{x_{j+c}})$$

and using the linearity of the real part of complex number. \square

We will now make use of Van der Corput's inequality for $H = \mathbb{C}$ to obtain another inequality:

Corollary 3.1 ([Assani, cor. 2.1])

For every finite sequence $u_0, u_1, \dots, u_{N-1} \in \mathbb{C}$ and integer $R \in \{0, 1, \dots, N-1\}$ the following inequality holds:

$$\sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n u_n \right|^2 \leq \frac{2}{N(R+1)} \sum_{n=0}^{N-1} |u_n|^2 + \frac{4}{R+1} \sum_{r=1}^R \left| \frac{1}{N} \sum_{n=0}^{N-r-1} u_n \overline{u_{n+r}} \right|.$$

Proof: Fix $\lambda \in \mathbb{T}$ and use Lemma 3.2 with $x_n := \lambda^n u_n$ to obtain

$$\begin{aligned} & \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n u_n \right|^2 \leq \\ & \leq \frac{N+R}{N(R+1)} \left(\frac{1}{N} \sum_{n=0}^{N-1} |\lambda^n u_n|^2 + \frac{2}{N(R+1)} \sum_{c=1}^R (R-c+1) \operatorname{Re} \left(\sum_{j=0}^{N-c-1} \lambda^j u_j \overline{\lambda^{j+c} u_{j+c}} \right) \right) \leq \\ & \leq \frac{2N}{N(R+1)} \left(\frac{1}{N} \sum_{n=0}^{N-1} |u_n|^2 + \frac{2(R+1)}{N(R+1)} \sum_{c=1}^R \operatorname{Re} \left(\sum_{j=0}^{N-c-1} \lambda^j \lambda^{-j-c} u_j \overline{u_{j+c}} \right) \right) \leq \\ & \leq \frac{2}{N(R+1)} \sum_{n=0}^{N-1} |u_n|^2 + \frac{4}{N(R+1)} \sum_{c=1}^R \left| \lambda^{-c} \sum_{j=0}^{N-c-1} u_j \overline{u_{j+c}} \right| = \\ & = \frac{2}{N(R+1)} \sum_{n=0}^{N-1} |u_n|^2 + \frac{4}{R+1} \sum_{r=1}^R \left| \frac{1}{N} \sum_{n=0}^{N-r-1} u_n \overline{u_{n+r}} \right|. \end{aligned}$$

Since the right-hand side of the above inequality is independent from λ , we can take supremum over $\lambda \in \mathbb{T}$ to finish the proof. \square

Now we are ready to give the proof of the Bourgain's uniform Wiener-Wintner theorem.

Proof: (of the Theorem 3.2)

Let's fix $f \in \mathcal{K}^\perp$, $x \in X$ and consider the sequence $u_n := f(T^n x)$. From Corollary 3.1 we have

$$\sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right|^2 \leq \frac{2}{N(R+1)} \sum_{n=0}^{N-1} |f(T^n x)|^2 + \frac{4}{R+1} \sum_{r=1}^R \left| \frac{1}{N} \sum_{n=0}^{N-r-1} f(T^n x) \overline{f(T^{n+r} x)} \right|$$

for every $N \in \mathbb{N}$, $R \leq N-1$. By Birkhoff's Ergodic Theorem (note that $f \in L^2(\mu) \Rightarrow |f| \in L^1(\mu)$) we have

$$\begin{aligned} & \limsup_{N \rightarrow \infty} \sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right|^2 \leq \\ & \frac{2}{R+1} \limsup_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |f(T^n x)|^2 + \frac{4}{R+1} \sum_{r=1}^R \left| \limsup_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-r-1} f(T^n x) \overline{f(T^{n+r} x)} \right| = \\ & = \frac{2}{R+1} \int_X |f|^2 d\mu + \frac{4}{R+1} \sum_{r=1}^R \left| \int_X f \overline{f \circ T^r} d\mu \right| = \\ & = \frac{2}{R+1} \int_X |f|^2 d\mu + \frac{4}{R+1} \sum_{r=1}^R |\langle f, U_T^r f \rangle| = \frac{2}{R+1} \int_X |f|^2 d\mu + \frac{4}{R+1} \sum_{r=1}^R |\sigma_f(r)|, \end{aligned} \tag{3.6}$$

which is valid for every $R \in \mathbb{N}$. By Lemma 3.1 we know that the measure σ_f is continuous, so by Wiener's Criterion of Continuity (Corollary 1.5) we have

$$\begin{aligned} \lim_{R \rightarrow \infty} \frac{1}{R+1} \sum_{r=1}^R |\hat{\sigma}_f(r)| &= \lim_{R \rightarrow \infty} \frac{R}{R+1} \frac{1}{R} \sum_{r=0}^{R-1} |\hat{\sigma}_f(r)| + \lim_{R \rightarrow \infty} \frac{1}{R+1} (\hat{\sigma}_f(R) - \hat{\sigma}_f(0)) = \\ &= \lim_{R \rightarrow \infty} \frac{R}{R+1} \cdot \lim_{R \rightarrow \infty} \frac{1}{R} \sum_{r=0}^{R-1} |\hat{\sigma}_f(r)| = \lim_{R \rightarrow \infty} \frac{1}{R} \sum_{r=0}^{R-1} |\hat{\sigma}_f(r)| = 0, \end{aligned}$$

since by Cauchy-Schwarz inequality $|\hat{\sigma}_f(R)| = |\langle U_T^R f, f \rangle| \leq \|U_T^R f\|_2 \|f\|_2 = \|f\|_2^2$. By taking $\lim_{R \rightarrow \infty}$ on both sides of (3.6) (left side is independent from R) we get

$$\limsup_{N \rightarrow \infty} \sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right|^2 \leq \lim_{R \rightarrow \infty} \frac{2}{R+1} \int_X |f|^2 d\mu + \lim_{R \rightarrow \infty} \frac{4}{R+1} \sum_{r=1}^R |\hat{\sigma}_f(r)| = 0. \quad \square$$

It is worth noticing that Bourgain's uniform Wiener-Wintner Theorem can be strengthened:

Proposition 3.1 ([Assani, Presser, Theorem 1.12])

Let (X, \mathcal{A}, μ, T) be a measure preserving dynamical system and take $f \in L^2(\mu)$. If

$$\lim_{N \rightarrow \infty} \sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right| = 0 \quad \mu\text{-a.e.},$$

then $f \in \mathcal{K}^\perp$.

Proof: Take $f \in L^2(\mu)$ and assume that $\lim_{N \rightarrow \infty} \sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right| = 0 \quad \mu\text{-a.e.}$. Then for every $\lambda \in \mathbb{T}$ we have

$$0 \leq \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right| \leq \sup_{\lambda \in \mathbb{T}} \left| \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \right| \xrightarrow{N \rightarrow \infty} 0 \quad \mu\text{-a.e.},$$

so $\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \xrightarrow{N \rightarrow \infty} 0 \quad \mu\text{-a.e.}$ On the other hand, by Corollary 2.1 $\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \xrightarrow{L^2(\mu)} P_{\bar{\lambda}} f$. $L^2(\mu)$ and $\mu\text{-a.e.}$ limits are $\mu\text{-a.e.}$ equal (if they both exists), so we have $P_{\bar{\lambda}} f = 0$. Since $\lambda \in \mathbb{T}$ was arbitrary, f must be orthogonal to every eigenfunction of U_T , so $f \in \mathcal{K}^\perp$. \square

3.2. Proof of Wiener-Wintner Ergodic Theorem

In this section we will prove the Wiener-Wintner Ergodic Theorem using Theorem 3.2. In order to do that we need another lemma.

Lemma 3.3 ([Eisner et al, lemma 21.7])

Let (X, \mathcal{A}, μ, T) be an ergodic dynamical system and take $f, f_1, f_2, \dots \in L^1(\mu)$ such that $f_n \xrightarrow{L^1(\mu)} f$.

There exists a set $X_0 \in \mathcal{A}$ of full measure, such that for $x \in X_0$ the following property holds: if $(a_n)_{n \in \mathbb{N}_0}$ is a bounded sequence in \mathbb{C} and the limit

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} a_n f_j(T^n x)$$

exists for every $j \in \mathbb{N}$, then also exists the limit

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

Proof: Take as X_0 the set of all $x \in X$ such that the limits $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |f_j(T^n x)|$ and $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} |(f - f_j)(T^n x)|$ exists. By Birkhoff Ergodic Theorem ($f - f_j \in L^1(\mu)$) we have that $\mu(X_0) = 1$ (as a countable intersection of full measure sets on which there is convergence). Take a bounded sequence $(a_n)_{n \in \mathbb{N}_0}$ in \mathbb{C} and suppose that $x \in X_0$ is such that the limit $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} a_n f_j(T^n x) =: b_j$ exists. Since $f_n \xrightarrow{L^1(\mu)} f$, $(\|f_j\|_1)_{j \in \mathbb{N}}$ is bounded. Take $K = \sup_{j \in \mathbb{N}} \|f_j\|_1$ and $M = \sup_{n \in \mathbb{N}_0} |a_n|$. We have

$$|b_j| = \left| \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} a_n f_j(T^n x) \right| \leq \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} MK = MK,$$

so the sequence $(b_j)_{j \in \mathbb{N}}$ is also bounded. Hence it has a convergent subsequence $(b_{j_m})_{m \in \mathbb{N}}$ with $\lim_{m \rightarrow \infty} b_{j_m} =: b$. Fix $\varepsilon > 0$ and take $m \in \mathbb{N}$ large enough to have $|b_{j_m} - b| < \frac{\varepsilon}{2}$ and $\|f - f_{j_m}\|_1 < \frac{\varepsilon}{2M}$. Now we have

$$\begin{aligned} \left| \frac{1}{N} \sum_{n=0}^{N-1} a_n f(T^n x) - b \right| &\leq \left| \frac{1}{N} \sum_{n=0}^{N-1} a_n f(T^n x) - a_n f_{j_m}(T^n x) \right| + \left| \frac{1}{N} \sum_{n=0}^{N-1} a_n f_{j_m}(T^n x) - b_{j_m} \right| + |b_{j_m} - b| < \\ &< \frac{1}{N} \sum_{n=0}^{N-1} M |f - f_{j_m}|(T^n x) + \left| \frac{1}{N} \sum_{n=0}^{N-1} a_n f_{j_m}(T^n x) - b_{j_m} \right| + \frac{\varepsilon}{2}, \end{aligned}$$

hence

$$\begin{aligned} \limsup_{N \rightarrow \infty} \left| \frac{1}{N} \sum_{n=0}^{N-1} a_n f(T^n x) - b \right| &< \\ \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} M |f - f_{j_m}|(T^n x) + \lim_{N \rightarrow \infty} \left| \frac{1}{N} \sum_{n=0}^{N-1} a_n f_{j_m}(T^n x) - b_{j_m} \right| + \frac{\varepsilon}{2} &< \\ &< M \|f - f_{j_m}\|_1 + 0 + \frac{\varepsilon}{2} = M \cdot \frac{\varepsilon}{2M} + \frac{\varepsilon}{2} = \varepsilon. \end{aligned}$$

Finally, since $\varepsilon > 0$ was arbitrary, we have

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} a_n f(T^n x) = b,$$

what completes the proof. \square

Corollary 3.2

Let (X, \mathcal{A}, μ, T) be an ergodic dynamical system and take $f, f_1, f_2, \dots \in L^1(\mu)$ such that $f_n \xrightarrow{L^1(\mu)} f$. If every f_n has the Wiener-Wintner property, then f also has the Wiener-Wintner property.

Proof: Let X_0 be the set from Lemma 3.3 and for $j \in \mathbb{N}$ let $X_j \in \mathcal{A}$ be such that $\mu(X_j) = 1$ and for $x \in X_j$ the limit $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f_j(T^n x)$ exists for all $\lambda \in \mathbb{T}$. Take a set $A := X_0 \cap \bigcap_{j=1}^{\infty} X_j$ and note that $\mu(A) = 1$. Fix $\lambda \in \mathbb{T}$ and $x \in A$. For $j \in \mathbb{N}$ we have $x \in X_j$, so the limit $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f_j(T^n x)$ exists for all $j \in \mathbb{N}$. Moreover we have $x \in X_0$, hence the limit $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x)$ also exists for all $\lambda \in \mathbb{T}$ (by the Lemma 3.3 with $a_n := \lambda^n$ (note that $|\lambda^n| \leq 1$)). \square

Proof: (of the Theorem 3.1)

First let's take $f \in L^2(\mu)$ being an eigenfunction of the Koopman operator U_T , i.e. suppose that there exists $\omega \in \mathbb{T}$ such that $f \circ T = \omega f$ μ -a.e. For almost all $x \in X$ and all $\lambda \in \mathbb{T}$ we have

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) = \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n \omega^n f(x) = f(x) \frac{1}{N} \sum_{n=0}^{N-1} (\lambda \omega)^n \xrightarrow{N \rightarrow \infty} f(x) \mathbb{1}_{\{1\}}(\lambda \omega),$$

so f has the Wiener-Wintner property. Now consider $f = \alpha_1 f_1 + \alpha_2 f_2 + \dots + \alpha_m f_m$, where $m \in \mathbb{N}$, $\alpha_1, \dots, \alpha_m \in \mathbb{C}$ and $f_1, \dots, f_m \in L^2(\mu)$ have the Wiener-Wiener property. Since

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) = \sum_{j=1}^m \alpha_j \left(\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f_j(T^n x) \right),$$

the limit $\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x)$ exists for each $x \in X_{f_1} \cap X_{f_2} \cap \dots \cap X_{f_m}$, so f has the Wiener-Wintner property (the set of all functions having Wiener-Wintner property forms a linear subspace of $L^1(\mu)$). Further, Corollary 3.2 shows that the set of all functions having Wiener-Wintner property is in fact a closed subspace of $L^1(\mu)$. Since we already know that eigenfunctions have the Wiener-Wiener property, then also every f from the Kronecker Factor \mathcal{K} must have the Wiener-Wintner property (since if $L^2(\mu) \ni f_j \xrightarrow{L^2(\mu)} f \in L^2(\mu)$ then also $f_j \xrightarrow{L^1(\mu)} f$, so $\mathcal{K} \subset \overline{\text{span}}^{L^1(\mu)} \{f \in L^2(\mu) : f \circ T = \lambda f \text{ for some } \lambda \in \mathbb{C}\}$). Observe further, that Theorem 3.2 implies that every $f \in \mathcal{K}^\perp$ has the Wiener-Wintner property (pointwise convergence for every $\lambda \in \mathbb{T}$ is weaker notion than uniform convergence for $\lambda \in \mathbb{T}$). By the Orthogonal Projection Theorem we have $L^2(\mu) = \mathcal{K} \oplus \mathcal{K}^\perp$, so since both \mathcal{K} and \mathcal{K}^\perp have the Wiener-Wintner property and the property is additive, the whole $L^2(\mu)$ has the Wiener-Wintner property. Since $L^2(\mu)$ is dense in $L^1(\mu)$ the proof is complete. \square

3.3. Uniform topological Wiener-Wintner theorem

In this section we will prove Wiener-Wintner theorem for topological dynamical systems, which is a combination of Wiener-Wintner Ergodic Theorem and Theorem 2.4. Both the theorem and

its proof are taken from [Robinson]. Following the author, we consider the case when X is a compact metric space, although the Theorem 5.1 shows that it's true also for nonmetrizable compact spaces. Let us begin with some facts about eigenfunctions of Koopman Operators on $C(X)$ and $L^2(\mu)$.

Let (X, T) be a uniquely ergodic topological dynamical system with unique ergodic measure μ . Denote by $M_T \subset \mathbb{C}$ the set of all eigenvalues of the Koopman Operator $U_T : L^2(X, \mathcal{B}(X), \mu) \rightarrow L^2(X, \mathcal{B}(X), \mu)$. By Proposition 2.2 we have $M_T \subset \mathbb{T}$. Since T is continuous, we can consider U_T as a Koopman Operator on $C(X)$. Denote by C_T the set of "continuous" eigenvalues from \mathbb{T} , i.e. $C_T = \{\lambda \in \mathbb{T} : f \circ T = \lambda f \text{ for some } f \in C(X) \setminus \{0\}\}$. We say that $f \in C(X) \setminus \{0\}$ with $f(Tx) = \lambda f(x)$ for all $x \in X$ is a *continuous eigenfunction* and $f \in L^2(\mu) \setminus \{0\}$ with $f \circ T = \lambda f$ μ -a.e. is a *measurable eigenfunction* (so continuous eigenfunctions are eigenvectors for Koopman Operator on $C(X)$ and measurable eigenfunctions are eigenvectors for Koopman Operator on $L^2(\mu)$). We know that measurable eigenfunction f satisfies $|f| = \text{const}$ μ -a.e. We will show that for continuous eigenfunction $|f(x)| = \text{const}$ for all $x \in X$.

Lemma 3.4

Let (X, T) be a uniquely ergodic topological dynamical system with unique ergodic probability measure μ . Suppose that $A \subset X$ is closed with $T(A) \subset A$. If $\text{supp}(\mu) \cap A = \emptyset$ then $A = \emptyset$.

Proof: Suppose that $A \neq \emptyset$. Consider topological dynamical system $(A, T|_A)$ (A is compact, since is a closed subset of a compact space). By Theorem 2.3 there exists regular Borel probability ν_0 on $(A, \mathcal{B}(A))$ which is $T|_A$ -invariant. Define a measure $\nu(B) := \nu_0(B \cap A)$ for $B \in \mathcal{B}(X)$. ν is a regular Borel probability. We will show that ν is T -invariant. For $B \in \mathcal{B}(X)$

$$\begin{aligned} \nu(T^{-1}B) &= \nu_0(T^{-1}B \cap A) = \nu_0(T|_A^{-1}(B) \cap A) = \nu_0(T|_A^{-1}(B) \cap T|_A^{-1}(A)) = \\ &= \nu_0(T|_A^{-1}(B \cap A)) = \nu_0(B \cap A) = \nu(B). \end{aligned}$$

Now ν is a T -invariant Borel probability measure on X with $\nu(A) = 1$. There must be $\mu \neq \nu$, since $\mu(A) \leq \mu(X \setminus \text{supp}(\mu)) = 0$. This is a contradiction with uniqueness of μ . \square

Lemma 3.5

Let (X, T) be a uniquely topological dynamical system with unique ergodic probability measure μ . Every continuous eigenfunction corresponding to eigenvalue $\lambda \in \mathbb{T}$ has constant modulus, i.e. there is $c \geq 0$ with $|g(x)| = c$ for all $x \in X$.

Proof: We have $|g| \circ T = |g \circ T| = |\lambda g| = |\lambda||g| = |g|$ and since T is ergodic we have by Proposition 2.1 $|g| = c$ μ -a.e. for some $c \geq 0$. the set $\{|g| \neq c\}$ is open (since g is continuous) and has measure zero, so $\text{supp}(\mu) \cap \{|g| \neq c\} = \emptyset$ (any point in $\{|g| \neq c\}$ has open neighborhood with zero measure, so cannot be in the support of μ), hence for any $x \in \text{supp}(\mu)$ there is $|g(x)| = c$. Now for any $d \geq 0$ with $d \neq c$ the set $\{|g| = d\}$ is closed (since T is continuous) with $T(\{|g| = d\}) \subset \{|g| = d\}$ (since $|g(x)| = d \Rightarrow |g(Tx)| = |\lambda||g(x)| = d$) and $\{|g| = d\} \subset \{|g| \neq c\}$, hence $\text{supp}(\mu) \cap \{|g| = d\} = \emptyset$. By Lemma 3.4 there must be $\{|g| = d\} = \emptyset$. This proves that $\{|g| = c\} = X$. \square

Corollary 3.3

Let (X, T) be a uniquely topological dynamical system with unique ergodic probability measure μ . Then $C_T \cap \mathbb{T} \subset M_T$.

Proof: Take $\lambda \in C_T \cap \mathbb{T}$ and $g \in C(X) \setminus 0$ with $g \circ T = \lambda g$. Then also $g \in L^2(\mu)$ and $g \circ T = \lambda g$ μ -a.e. It only remains to show that $g \neq 0$ in $L^2(\mu)$, that is $\mu(\{g \neq 0\}) > 0$ (note that in general $g \neq 0$ in $C(X)$ doesn't imply that $g \neq 0$ in $L^2(\mu)$ - g can be nonzero on set of measure zero). From Lemma 3.5 we have $|g| = \text{const}$ and since $g \neq 0$ in at least one point, $|g(x)| \neq 0$ for all $x \in X$, hence also $\mu(\{g \neq 0\}) = \mu(X) > 0$. \square

It follows from the Proposition 2.2 that for ergodic T , the eigenspace corresponding to the eigenvalue $\lambda \in M_T$ is one-dimensional, so we can explicitly give form of the orthogonal projection on this eigenspace. We have

$$P_\lambda f = \frac{\langle f, g \rangle}{\|g\|_2^2} g \text{ for } f \in L^2(\mu), \quad (3.7)$$

where g is an eigenfunction corresponding to λ . If $\lambda \in \mathbb{T} \setminus M_T$, then $P_\lambda f = 0$ for all $f \in L^2(\mu)$. For $\lambda \in C_T$ we can claim that g is a continuous function and (3.7) defines an operator $P_\lambda : C(X) \rightarrow C(X)$ and $P_\lambda f = 0$ for $\lambda \notin (C_T \cup M_T)$. We will not define $P_\lambda : C(X) \rightarrow C(X)$ for $\lambda \in M_T \setminus C_T$.

Observe that $\lambda \in M_T \Leftrightarrow \lambda^{-1} = \bar{\lambda} \in M_T$ and $\lambda \in C_T \Leftrightarrow \bar{\lambda} \in C_T$. It follows from facts that $g(Tx) = \lambda g(x) \Leftrightarrow \bar{g}(Tx) = \bar{\lambda} \bar{g}(x)$; $g \neq 0 \Leftrightarrow \bar{g} \neq 0$ and g is continuous if and only if \bar{g} is continuous. It follows that for $\lambda \in \mathbb{T}$ with $\lambda \notin M_T \setminus C_T$ we also have $\bar{\lambda} \notin M_T \setminus C_T$. Now we state uniform Wiener-Wintner theorem for topological dynamical systems.

Theorem 3.3 (Uniform topological Wiener-Wintner Theorem, [Robinson, thm 1.1])

Let (X, T) be a uniquely ergodic topological dynamical system on a compact metric space X with unique ergodic probability measure μ . Then for all $\lambda \in \mathbb{T}$ with $\lambda \notin M_T \setminus C_T$ and $f \in C(X)$ the sequence

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x)$$

converges uniformly across $x \in X$ to $P_\lambda f \in C(X)$.

Remark Our formulation of the theorem is slightly different from the original one in [Robinson].

The author states that for the same λ we have $\frac{1}{N} \sum_{n=0}^{N-1} \lambda^{-n} f(T^n x) \xrightarrow{\|\cdot\|_{\sup}} P_\lambda f$. The observation made before the theorem implies that these two formulations are equivalent.

Since $f \in C(X)$ can be seen also as an element of $L^2(\mu)$, we can consider the spectral measure $\sigma_{f,T,\mu} := \sigma_{f,U_T}$ for the element f and Koopman operator $U_T : L^2(\mu) \rightarrow L^2(\mu)$.

Lemma 3.6 ([Robinson, lemma 2.1])

Let (X, T) be a uniquely ergodic topological dynamical system on a compact metric space X with unique ergodic probability measure μ . Let $(x_n)_{n \in \mathbb{N}}$ be a sequence in X . Then for all $f \in C(X)$ and $\lambda \in \mathbb{T}$ we have

$$\sigma_{f,T,\mu}(\{\bar{\lambda}\})^{1/2} \geq \limsup_{N \rightarrow \infty} \frac{1}{N} \left| \sum_{n=0}^{N-1} \lambda^n f(T^n x_N) \right|. \quad (3.8)$$

Proof: Take a subsequence $(x_{N_j})_{j \in \mathbb{N}}$ such that

$$\lim_{j \rightarrow \infty} \frac{1}{N_j} \left| \sum_{n=0}^{N_j-1} \lambda^n f(T^n x_{N_j}) \right| = \limsup_{N \rightarrow \infty} \frac{1}{N} \left| \sum_{n=0}^{N-1} \lambda^n f(T^n x_N) \right|. \quad (3.9)$$

Consider topological dynamical system (\hat{X}, \hat{T}) , where $\hat{X} = X \times \mathbb{T}$ and $\hat{T} = T \times R_\lambda$, i.e. $\hat{T}(x, \omega) = (Tx, \lambda\omega)$. Note that \hat{X} is a metric compact space. By Fact 2.2, $\mu \otimes m$ is \hat{T} -invariant. Consider the sequence of measures $\nu_{N_j} = \delta_{(x_{N_j}, 1)} \in \mathcal{M}^1(\hat{X})$. By Banach-Alaoglu Theorem there exists measure $\rho \in \mathcal{M}^1(\hat{X})$ with $\frac{1}{N_j} \sum_{n=0}^{N_j} \hat{T}_*^n \nu_{N_j} = \frac{1}{N_j} \sum_{n=0}^{N_j} \delta_{(T^n x_{N_j}, \lambda^n)} \xrightarrow{*w} \rho$ (possibly after passing to subsequence). By Lemma 2.1 ρ is \hat{T} -invariant. Take function $h \in C(\hat{X})$ given by $h(x, \omega) = f(x)\omega$. We have

$$\frac{1}{N_j} \sum_{n=0}^{N_j} f(T^n x_{N_j}) \lambda^n = \frac{1}{N_j} \sum_{n=0}^{N_j} \int_{\hat{X}} h d\delta_{(T^n x_{N_j}, \lambda^n)} \longrightarrow \int_{\hat{X}} h d\rho, \quad (3.10)$$

since $\frac{1}{N_j} \sum_{n=0}^{N_j} \delta_{(T^n x_{N_j}, \lambda^n)} \xrightarrow{*w} \rho$. The marginal measure $\rho|_X : \mathcal{B}(X) \rightarrow [0, 1]$ defined by $\rho|_X(B) = \rho(B \times \mathbb{T})$ for $B \in \mathcal{B}(X)$ is a probability measure on $(X, \mathcal{B}(X))$. Now we will verify that $\rho|_X$ is also T -invariant:

$$\rho|_X(T^{-1}B) = \rho(T^{-1}(B) \times \mathbb{T}) = \rho(\hat{T}^{-1}(B \times \mathbb{T})) = \rho(B \times \mathbb{T}) = \rho|_X(B).$$

Since (X, T) is uniquely ergodic $\rho|_X = \mu$. Let's take spectral measure $\sigma_{h, \hat{T}, \rho}$ of the function h for the Koopman Operator $U_{\hat{T}} : L^2(\rho) \rightarrow L^2(\rho)$. Observe that

$$\begin{aligned} \widehat{\sigma}_{h, \hat{T}, \rho}(n) &= \langle U_{\hat{T}} h, h \rangle = \int_{\hat{X}} h \circ \hat{T}^n \bar{h} d\rho = \int_{\hat{X}} h(T^n x, \lambda^n \omega) \overline{h(x, \omega)} d\rho(x, \omega) = \\ &= \int_{\hat{X}} f(T^n x) \lambda^n \omega \overline{f(x) \omega} d\rho(x, \omega) = \lambda^n \int_{\hat{X}} f(T^n x) \overline{f(x)} d\rho(x, \omega) = \\ &= \lambda^n \int_X f(T^n x) \overline{f(x)} d\rho|_X(x) = \lambda^n \int_X f(T^n x) \overline{f(x)} d\mu(x) = \lambda^n \widehat{\sigma}_{f, T, \mu}(n). \end{aligned}$$

Consider the measure $\sigma_{h, \hat{T}, \rho} \circ R_\lambda^{-1}$ on $(\mathbb{T}, \mathcal{B}(\mathbb{T}))$ given by $\sigma_{h, \hat{T}, \rho} \circ R_\lambda^{-1}(A) := \sigma_{h, \hat{T}, \rho}(R_\lambda^{-1}(A))$ (so it's measure $\sigma_{h, \hat{T}, \rho}$ transported by rotation R_λ). We have

$$\begin{aligned} \widehat{\sigma_{h, \hat{T}, \rho} \circ R_\lambda^{-1}}(n) &= \int_{\mathbb{T}} z^n d\sigma_{h, \hat{T}, \rho} \circ R_\lambda^{-1}(z) = \int_{\mathbb{T}} (\bar{\lambda} z)^n d\sigma_{h, \hat{T}, \rho}(z) = \\ &= \lambda^{-n} \int_{\mathbb{T}} z^n d\sigma_{h, \hat{T}, \rho}(z) = \lambda^{-n} \widehat{\sigma}_{h, \hat{T}, \rho}(n) = \lambda^{-n} \lambda^n \widehat{\sigma}_{f, T, \mu}(n) = \widehat{\sigma}_{f, T, \mu}(n). \end{aligned}$$

By Fourier Uniqueness Theorem (Theorem 1.13) $\sigma_{h, \hat{T}, \rho} \circ R_\lambda^{-1} = \sigma_{f, T, \mu}$. In particular

$$\sigma_{f, T, \mu}(\{\bar{\lambda}\}) = \sigma_{h, \hat{T}, \rho} \circ R_\lambda^{-1}(\{\bar{\lambda}\}) = \sigma_{h, \hat{T}, \rho}(R_\lambda^{-1}(\{\bar{\lambda}\})) = \sigma_{h, \hat{T}, \rho}(\{1\}). \quad (3.11)$$

From Lemma 2.2 we know that

$$\sigma_{h, \hat{T}, \rho}(\{1\}) = \|\hat{P}_1 h\|_2^2, \quad (3.12)$$

where \hat{P}_1 is an orthogonal projection on the eigenspace $H_1 := \{f \in L^2(\rho) : f = f \circ \hat{T}\}$ of 1 for the Koopman Operator $U_{\hat{T}} : L^2(\rho) \rightarrow L^2(\rho)$. Note that $H_{\text{const}} := \{f \in L^2(\rho) : f = \text{const } \rho\text{-a.e.}\}$

is a closed linear subspace in $L^2(\rho)$ and denote the orthogonal projection on the subspace H_{const} as \hat{P}_{const} . There is $H_{\text{const}} \subset H_1$, hence

$$\|P_{\text{const}}f\|_2 \leq \|P_1f\|_2 \text{ for all } f \in L^2(\rho). \quad (3.13)$$

On the other hand, the subspace H_{const} is one-dimensional with $1 \in H_{\text{const}}$, so $P_{\text{const}}h = \langle h, 1 \rangle \cdot 1 = \int_{\hat{X}} h d\rho$. Combinig this fact with (3.9), (3.10), (3.11), (3.12) and (3.13) we get

$$\begin{aligned} \sigma_{f,T,\mu}(\{\bar{\lambda}\})^{1/2} &= \sigma_{h,\hat{T},\rho}(\{1\})^{1/2} = \|\hat{P}_1h\|_2 \geq \|\hat{P}_{\text{const}}h\|_2 = \left| \int_{\hat{X}} h d\rho \right| = \\ &= \lim_{j \rightarrow \infty} \left| \frac{1}{N_j} \sum_{n=0}^{N_j} f(T^n x_{N_j}) \lambda^n \right| = \limsup_{N \rightarrow \infty} \frac{1}{N} \left| \sum_{n=0}^{N-1} \lambda^n f(T^n x_N) \right|. \quad \square \end{aligned}$$

With the above lemma in hand we are ready to give proof of the Theorem 3.3.

Proof: (of the Theorem 3.3)

Take $\lambda \in \mathbb{T}$ with $\lambda \notin M_T \setminus C_T$ and $f \in C(X)$.

- (1) Suppose that $\lambda \in C_T$. There exists $g \in C(X) \setminus \{0\}$ with $g \circ T = \lambda g$ and also $\bar{g} \circ T = \bar{\lambda} \bar{g}$. Take $h = fg \in C(X)$. By Theorem 2.4

$$\begin{aligned} \frac{1}{N} \sum_{n=0}^{N-1} h(T^n x) &= \frac{1}{N} \sum_{n=0}^{N-1} f(T^n x) g(T^n x) = \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) g(x) \xrightarrow{\|\cdot\|_{\text{sup}}} \int_X h d\mu = \\ &= \int_X f g d\mu = \langle f, \bar{g} \rangle. \end{aligned} \quad (3.14)$$

After multiplying both sides of (3.14) by $\bar{g}(x)$ we get

$$|g(x)|^2 \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x) \xrightarrow{\|\cdot\|_{\text{sup}}} \langle f, \bar{g} \rangle \bar{g}(x).$$

Applying Lemma 3.5 the function $|g|$ is constant (and nonzero), so $|g(x)|^2 = \int_X |g|^2 d\mu = \|g\|_2^2$ for all $x \in X$ and further by (3.7)

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f \circ T^n \xrightarrow{\|\cdot\|_{\text{sup}}} \frac{\langle f, \bar{g} \rangle}{\|g\|_2^2} \bar{g} = P_{\bar{\lambda}} f.$$

- (2) Now let us discuss the case $\lambda \notin M_T$. Then $P_{\bar{\lambda}} f = 0$ and we have to show that $\|\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f \circ T^n\|_{\text{sup}} \rightarrow 0$. Suppose on the contrary, that there exists $\varepsilon > 0$ and a sequence $(x_n)_{n \in \mathbb{N}}$ such that $|\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f(T^n x_N)| \geq \varepsilon$ for all $N \in \mathbb{N}$. By Lemma 3.6 we have

$$\sigma_{f,T,\mu}(\{\bar{\lambda}\})^{1/2} \geq \limsup_{N \rightarrow \infty} \frac{1}{N} \left| \sum_{n=0}^{N-1} \lambda^n f(T^n x_N) \right| \geq \varepsilon > 0.$$

On the other hand, from Lemma 2.2 we know that

$$\sigma_{f,T,\mu}(\{\bar{\lambda}\})^{1/2} = \|P_{\bar{\lambda}}f\|_2 = 0.$$

Obtained contrary finishes the proof. \square

It's also possible to obtain equivalence in the Theorem 3.3.

Proposition 3.2

Let (X, T) be a uniquely ergodic topological dynamical system with unique measure μ and take $\lambda \in \mathbb{T}$. If for every $f \in C(X)$ the sequence

$$\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f \circ T^n \quad (3.15)$$

converges uniformly, then $\lambda \notin M_T \setminus C_T$.

Proof: If $\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f \circ T^n \xrightarrow{\|\cdot\|_{\sup}} Qf$ for all $f \in C(X)$, then also $\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f \circ T^n \xrightarrow{L^2(\mu)} Qf$. By Corollary 2.1 $Q = P_{\bar{\lambda}}$ on $C(X)$. If $P_{\bar{\lambda}}f \neq 0$ for some $f \in C(X)$ then $P_{\bar{\lambda}}f = Qf$ is a continuous eigenfunction for $\bar{\lambda}$ (it is easy to verify that Qf satisfies $Qf \circ T = \bar{\lambda}Qf$ pointwise), hence $\lambda \in C_T$. Otherwise $P_{\bar{\lambda}} = 0$ on a dense subset $C(X) \subset L^2(\mu)$, hence $P_{\bar{\lambda}} = 0$ on $L^2(\mu)$ and $\lambda \notin M_T$. \square

Let us finish this chapter with an example of the uniquely ergodic topological dynamical system (X, T) for which there exists $\lambda \in \mathbb{T}$ and continuous function $f \in C(X)$ such that the sequence $\frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f \circ T^n$ does not converge uniformly. The example is due to Roland Derndinger.

Example 3.1 (R. Derndinger, [Schreiber14, ex. 2.4]) Let us consider the metric space $\{-1, 1\}^{\mathbb{N}}$, i.e. the set of all sequences $\underline{x} : \mathbb{N} \rightarrow \{-1, 1\}$ with the metric $d : \{-1, 1\}^{\mathbb{N}} \times \{-1, 1\}^{\mathbb{N}} \rightarrow [0, \infty)$ defined by $d(\underline{x}, \underline{y}) = \sum_{n=1}^{\infty} \frac{|x_n - y_n|}{2^n}$ for $\underline{x} = (x_n)_{n \in \mathbb{N}} \in \{-1, 1\}^{\mathbb{N}}$, $\underline{y} = (y_n)_{n \in \mathbb{N}} \in \{-1, 1\}^{\mathbb{N}}$. The Tychonoff's theorem yields that $\{-1, 1\}^{\mathbb{N}}$ is a compact space. Define the sequence

$$\underline{x}^{(i)} = (x_n^{(i)})_{n \in \mathbb{N}}, \text{ where } i \in \mathbb{N} \text{ and } x_n^{(i)} := (-1)^n \mathbb{1}_{\{1, \dots, i-1\}}(n) + (-1)^{n+1} \mathbb{1}_{\{i, i+1, \dots\}}(n) \text{ for } n \in \mathbb{N}.$$

In particular there is

$$\begin{aligned} \underline{x}^{(1)} &= (1, -1, 1, -1, 1, -1, \dots), \quad \underline{x}^{(2)} = (-1, -1, 1, -1, 1, -1, \dots), \\ \underline{x}^{(3)} &= (-1, 1, 1, -1, 1, -1, \dots), \quad \underline{x}^{(4)} = (-1, 1, -1, -1, 1, -1, \dots) \text{ and so on.} \end{aligned}$$

The set $K := \{\pm \underline{x}^{(i)} : i \in \mathbb{N}\}$ is a compact subspace of $\{-1, 1\}^{\mathbb{N}}$. Indeed, let us take a sequence $(\underline{y}(m))_{m \in \mathbb{N}}$ from K with $\underline{y}(m) \rightarrow \underline{x} \in \{-1, 1\}^{\mathbb{N}}$. For every $m \in \mathbb{N}$ there is $\underline{y}(m) = s_m \underline{x}^{(i_m)}$ for some $s_m \in \{-1, 1\}$ and $i_m \in \mathbb{N}$. Assume first that there exists $i \in \mathbb{N}$ with $i_m = i$ for infinitely many $m \in \mathbb{N}$. There exists also $s \in \{-1, 1\}$ such that $\underline{y}(m) = s \underline{x}^{(i)}$ for infinitely many $m \in \mathbb{N}$, hence there is $\underline{x} = s \underline{x}^{(i)} \in K$. Assume now that for any $i \in \mathbb{N}$ there is $i_m = i$ for only finitely many $m \in \mathbb{N}$. We can find subsequence $(i_{m_k})_{k \in \mathbb{N}}$ with $i_{m_k} \geq k$ for every $k \in \mathbb{N}$. The sequence $(i_{m_k})_{k \in \mathbb{N}}$ can be taken such that there is $s_{m_k} = s$ for some $s \in \{-1, 1\}$ and every $k \in \mathbb{N}$ (possibly after passing to subsequence). We will show that $\underline{y}(m_k) \rightarrow -s \underline{x}^{(1)}$. There is

$$d(\underline{y}(m_k), -s \underline{x}^{(1)}) = \sum_{n=1}^{\infty} \frac{|\underline{y}(m_k)_n - (-s x_n^{(1)})|}{2^n} = \sum_{n=1}^{\infty} \frac{|s \underline{x}_n^{(i_{m_k})} + s x_n^{(1)}|}{2^n} =$$

$$= \sum_{n=1}^{i_{m_k}-1} \frac{|(-1)^{n+1} + (-1)^n|}{2^n} + \sum_{n=i_{m_k}}^{\infty} \frac{|(-1)^n + (-1)^n|}{2^n} \leq \sum_{n=k}^{\infty} \frac{2}{2^n} = \frac{1}{2^{k-2}} \rightarrow 0.$$

Since $\underline{y}(m) \rightarrow \underline{x}$, there is $\underline{x} = -s\underline{x}^{(1)} \in K$. We proved that K is a closed subset of $\{-1, 1\}^{\mathbb{N}}$, hence is a compact space.

We will consider the shift transformation $T : K \rightarrow K$ given by $(T\underline{x})_n := (\underline{x})_{n+1}$. It is easy to verify that T is a continuous transformation. We will prove that T is uniquely ergodic on K . By the Theorem 2.4 it suffices to show that for every $f \in C(K)$ there exists constant

$c(f) \in \mathbb{C}$ with $\frac{1}{N} \sum_{n=0}^{N-1} f(T^n(\underline{x})) \rightarrow c(f)$ for every $\underline{x} \in K$. Note that for every $i \in \mathbb{N}$ there is $T^n(\pm \underline{x}^{(i)}) = \pm(-1)^n \underline{x}^{(1)}$ for $n > i$, hence $f(T^n(\pm \underline{x}^{(i)})) = f(\pm(-1)^n \underline{x}^{(1)})$ for $n \in \mathbb{N}$ large enough.

We get $\frac{1}{N} \sum_{n=0}^{N-1} f(T^n(\underline{x})) \rightarrow \frac{1}{2}(f(\underline{x}^{(1)}) + f(-\underline{x}^{(1)})) =: c(f)$, hence T is uniquely ergodic. This yields that the unique invariant probability measure μ is $\frac{1}{2}(\delta_{\underline{x}^{(1)}} + \delta_{-\underline{x}^{(1)}})$, since $c(f) = \int_K f d\mu$ for every

$f \in C(K)$. Let us take now $\lambda = -1 \in \mathbb{T}$ and $f \in C(K)$ given by $f(\underline{x}) = f((x_1, x_2, \dots)) = x_1$. For every $i \in \mathbb{N}$ there is $f(T^n(\pm \underline{x}^{(i)})) = f(\pm(-1)^n \underline{x}^{(1)}) = \pm(-1)^n$ for $n > i$, so $(-1)^n f(T^n(\pm \underline{x}^{(i)})) = (-1)^n \pm (-1)^n = \pm 1$. We have now

$$\frac{1}{N} \sum_{n=0}^{N-1} (-1)^n f(T^n(\pm \underline{x}^{(i)})) \rightarrow \pm 1 =: h(\pm \underline{x}^{(i)}),$$

for every $i \in \mathbb{N}$. Simillary as before, there is $\underline{x}^{(i)} \xrightarrow{i \rightarrow \infty} -\underline{x}^{(1)}$ and $1 = h(\underline{x}^{(i)}) \nrightarrow h(-\underline{x}^{(1)}) = -1$, hence h is not continuous on K . This proves that $\frac{1}{N} \sum_{n=0}^{N-1} (-1)^n f \circ T^n$ does not converge in $C(K)$. ■

Chapter 4

Ergodic theory for operator semigroups and Markov operators on $C(K)$

This chapter contains preliminaries for the next chapter in which we will establish topological Wiener-Winter theorem for semigroups of operators. We present basics of operator theoretic approach to ergodic theory. Instead of single deterministic transformation on measure space, we will consider semigroup of bounded linear operators on Banach space. Introduction to these topics can be found in [Krengel, Brunel]. We will follow setting presented in [Schreiber13] and [Schreiber14]. We also introduce concepts of amenable semigroups and Markov operators on $C(K)$.

4.1. Operator semigroups, ergodic nets and mean ergodicity

Definition 4.1 We say that semigroup G is a **semitopological semigroup** if G is endowed with Hausdorff topology, which makes transformations $G \ni g \mapsto gh \in G$, $G \ni g \mapsto hg \in G$ continuous for every $h \in G$.

Let \mathfrak{X} be a Banach space. We will consider semigroups of bounded operators on \mathfrak{X} indexed by elements of semitopological semigroup. We will assume such representation to have some basic properties. Space $L(\mathfrak{X})$ will be considered with strong operator topology, unless stated otherwise.

Definition 4.2 Let G be a semitopological semigroup. $\mathcal{S} = \{S_g : g \in G\}$ is called a **bounded representation of G on \mathfrak{X}** if

- (i) $S_g \in L(\mathfrak{X})$ for every $g \in G$,
- (ii) $S_{g_1}S_{g_2} = S_{g_2g_1}$ for all $g_1, g_2 \in G$,
- (iii) $\sup_{g \in G} \|S_g\| < \infty$,
- (iv) $G \ni g \mapsto S_g x$ is continuous for every $x \in \mathfrak{X}$.

For $x \in \mathfrak{X}$ we will denote $\mathcal{S}x := \{S_g x : g \in G\}$.

Example 4.1 Let X be a compact Hausdorff space, let $T : X \rightarrow X$ be a continuous transformation and $U_T \in L(C(X))$ be its Koopman operator on the Banach space of continuous functions. Note that U_T is a contraction, i.e. $\|U_T\| \leq 1$. Semigroup \mathbb{N} endowed with the discrete topology becomes semitopological semigroup. $\mathcal{S} := \{U_T^n : n \in \mathbb{N}\}$ is a bounded representation of \mathbb{N} on $C(X)$. Analogously, if $T : X \rightarrow X$ is a measure preserving transformation on the measure space (X, \mathcal{A}, μ) , then the Koopman operator $U_T : L^p(\mu) \rightarrow L^p(\mu), p \in [1, \infty)$ gives rise to a bounded representation $\{U_T^n\}_{n \in \mathbb{N}}$ of \mathbb{N} on $L^p(\mu)$. ■

Example 4.2 The previous example can be generalised. Let \mathfrak{X} be a Banach space and let $\{T_n\}_{n \in \mathbb{N}}$ be a sequence of bounded linear operators on \mathfrak{X} satisfying $T_{n+m} = T_m T_n$ for $n, m \in \mathbb{N}$ and assume that sequence $\{T_n\}_{n \in \mathbb{N}}$ is **power bounded**, i.e. $\sup_{n \in \mathbb{N}} \|T_n\| < \infty$. Then $\mathcal{S} := \{T_n : n \in \mathbb{N}\}$ is a bounded representation of \mathbb{N} on \mathfrak{X} . In particular, if we take T to be a contraction on \mathfrak{X} , then a sequence $T_n := T^n$ is a bounded representation of \mathbb{N} on \mathfrak{X} . ■

Note that bounded representation \mathcal{S} is itself a semitopological semigroup with respect to the strong operator topology. In fact, even $co\mathcal{S}$ and $\overline{co}\mathcal{S}$ are semitopological semigroups. We will denote fixed space of semigroup \mathcal{S} by

$$Fix(\mathcal{S}) := \{x \in \mathfrak{X} : S_g x = x \text{ for all } g \in G\}.$$

$Fix(\mathcal{S})$ is a closed linear subspace of \mathfrak{X} . We will also denote

$$\mathbb{P}_{\mathcal{S}} = \{\mu \in Fix(\mathcal{S}) : \mu \geq 0, \mu(X) = 1\}$$

for the set of invariant probability measures. Together with semigroup \mathcal{S} we will consider adjoint semigroup $\mathcal{S}' := \{S'_g : g \in G\}$ and its fixed space $Fix(\mathcal{S}')$.

For a given bounden representation \mathcal{S} , we would like to study its mean asymptotic behaviour, in a similar fashion in which we studied existence of limit for ergodic averages

$$\frac{1}{N} \sum_{n=0}^{N-1} f \circ T^n.$$

Since there is no canonical way of averaging for abstract semigroup, we need to introduce the concept of \mathcal{S} -ergodic net.

Definition 4.3 Let \mathcal{S} be a bounded representation of semigroup G . A net $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$ of operators in $L(\mathfrak{X})$ is called a **strong right (left) \mathcal{S} -ergodic net** if

- (i) $A_\alpha^\mathcal{S} \in \overline{co}\mathcal{S}$ for all $\alpha \in \Lambda$ (closure is taken with respect to the strong operator topology),
- (ii) $(A_\alpha^\mathcal{S})$ is **strongly right (left) asymptotically invariant**, i.e.

$$\lim_{\alpha} (A_\alpha^\mathcal{S} x - A_\alpha^\mathcal{S} S_g x) = 0 \quad \left(\lim_{\alpha} (A_\alpha^\mathcal{S} x - S_g A_\alpha^\mathcal{S} x) = 0 \right).$$

The net $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$ is called a **weak right (left) \mathcal{S} -ergodic net** if condition (i) holds and $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$ is **weakly right (left) asymptotically invariant**, i.e. convergence in (ii) holds in weak topology on \mathfrak{X} . The net $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$ is called **strong (weak) \mathcal{S} -ergodic net** when it is a strong (weak) right and left \mathcal{S} -ergodic net. We say that representation \mathcal{S} **admits** strong (weak) right/left ergodic net if there exists at least one ergodic net with desired properties.

Example 4.3 Let $\mathcal{S} := \{T_n\}_{n \in \mathbb{N}}$ be such as in Example 4.2. The sequence of Ceasaro averages

$$A_N := \frac{1}{N} \sum_{n=0}^{N-1} T_n, \quad N \in \mathbb{N},$$

is a strong \mathcal{S} -ergodic net. Indeed, we have $A_N \in \text{co}\mathcal{S}$ for all $N \in \mathbb{N}$ and

$$\begin{aligned} \|A_N x - A_N T_k x\| &= \left\| \frac{1}{N} \left(\sum_{n=0}^{N-1} T_n x - \sum_{n=0}^{N-1} T_{k+n} x \right) \right\| = \left\| \frac{1}{N} \left(\sum_{n=0}^{k-1} T_n x - \sum_{n=N}^{N+k-1} T_n x \right) \right\| \leq \\ &\leq \frac{2k\|x\|}{N} \sup_{n \in \mathbb{N}} \|T_n\| \xrightarrow{N \rightarrow \infty} 0 \text{ for every } k \in \mathbb{N}. \end{aligned}$$

Analogously there is

$$\|A_N x - T_k A_N x\| \xrightarrow{N \rightarrow \infty} 0 \text{ for every } k \in \mathbb{N},$$

which shows that $(A_N)_{n \in \mathbb{N}}$ is a strong \mathcal{S} -ergodic net. Of course, $(A_N)_{n \in \mathbb{N}}$ is also weak \mathcal{S} -ergodic net. In particular, Ceasaro averages of iterates of Koopman Operator for continuous transformation on compact space X forms a strong ergodic net on the space $C(X)$. ■

The following fact might be surprising:

Theorem 4.1 ([Schreiber13, thm 1.5])

Let \mathcal{S} be a bounded representation of semitopological semigroup G on \mathfrak{X} . There exists strong (right) \mathcal{S} -ergodic net if and only if there exists weak (right) \mathcal{S} -ergodic net.

Operator ergodic theory is interested in convergence of net $A_\alpha^\mathcal{S} x$ for each $x \in \mathfrak{X}$. We will now introduce central theorem of the ergodic theory for operator semigroups which deals with that convergence. Our statement of the theorem is a combination of Theorem 1.7 from [Schreiber13] and Theorems 1.5 (p. 76) and 1.9 (p. 79) from [Krengel, Brunel].

Theorem 4.2 (Abstract mean ergodic theorem)

Let \mathcal{S} be a bounded representation of semitopological semigroup G which admits strong right ergodic net. The following conditions are equivalent:

- (1) $A_\alpha^\mathcal{S} x$ converges to a fixed point of \mathcal{S} for every $x \in \mathfrak{X}$ and some/every strong right \mathcal{S} -ergodic net $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$,
- (2) $A_\alpha^\mathcal{S} x$ converges weakly to a fixed point of \mathcal{S} for every $x \in \mathfrak{X}$ and some/every strong right \mathcal{S} -ergodic net $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$,
- (3) $A_\alpha^\mathcal{S} x$ converges weakly to a fixed point of \mathcal{S} for every $x \in \mathfrak{X}$ and some/every weak right \mathcal{S} -ergodic net $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$,
- (4) $A_\alpha^\mathcal{S} x$ has a weak cluster point in $\text{Fix}(\mathcal{S})$ for every $x \in \mathfrak{X}$ and some/every weak right \mathcal{S} -ergodic net $(A_\alpha^\mathcal{S})_{\alpha \in \Lambda}$,
- (5) $\overline{\text{co}}\mathcal{S} \cap \text{Fix}(\mathcal{S}) \neq \emptyset$ for all $x \in \mathfrak{X}$,
- (6) there exists operator $P \in \overline{\text{co}}\mathcal{S}$ such that $S_g P = P S_g = P^2 = P$ for every $g \in G$,
- (7) $\text{Fix}(\mathcal{S})$ separates $\text{Fix}(\mathcal{S}')$, (i.e. for every $\text{Fix}(\mathcal{S}') \ni x^* \neq 0$ there exists $x \in \text{Fix}(\mathcal{S})$ with $\langle x, x^* \rangle \neq 0$),

$$(8) \mathfrak{X} = \text{Fix}(\mathcal{S}) \oplus \overline{\text{span}} \text{rg}(I - \mathcal{S}) := \text{Fix}(\mathcal{S}) \oplus \overline{\text{span}} \{x - S_g x : x \in \mathfrak{X}, g \in G\}.$$

The strong (weak) limit $\lim_{\alpha} A_{\alpha}^{\mathcal{S}} x$ is equal for every strong (weak) right \mathcal{S} -ergodic net $(A_{\alpha}^{\mathcal{S}})_{\alpha \in \Lambda}$ and is also equal to Px . Operator P is a \mathcal{S} -absorbing projection onto $\text{Fix}(\mathcal{S})$ along $\overline{\text{span}} \text{rg}(I - \mathcal{S})$.

Remark Note that conditions (5) - (8) in the above theorem doesn't require the concept of \mathcal{S} -ergodic net. Nevertheless, proof of equivalence between them requires existence of at least one right \mathcal{S} -ergodic net. This is why we will introduce concept of amenable semigroup for which representations there always exists ergodic net. Motivated by the Abstract Mean Ergodic Theorem, we define mean ergodicity of semigroup \mathcal{S} in terms of absorbing projection, which itself doesn't require concept of ergodic net.

Definition 4.4 Let \mathcal{S} be a bounded representation of semitopological semigroup G . We say that \mathcal{S} is **mean ergodic** if $\overline{\text{co}}\mathcal{S}$ contains \mathcal{S} -absorbing projection (also called a zero element), i.e. there exists $P \in \overline{\text{co}}\mathcal{S}$ with $P^2 = S_g P = P S_g = P$ for all $g \in G$.

If \mathcal{S} admits even \mathcal{S} -ergodic net (which is simultaneously left and right asymptotically invariant), then the limit $Px = \lim A_{\alpha}^{\mathcal{S}} x$ is always a fixed point of \mathcal{S} (if exists). It allows us to give the following variant:

Theorem 4.3 (Abstract mean ergodic theorem)

Let \mathcal{S} be a bounded representation of semitopological semigroup G which admits strong ergodic net. The following conditions are equivalent:

- (1) $A_{\alpha}^{\mathcal{S}} x$ converges for every $x \in \mathfrak{X}$ and some/every strong right \mathcal{S} -ergodic net $(A_{\alpha}^{\mathcal{S}})_{\alpha \in \Lambda}$,
- (2) $A_{\alpha}^{\mathcal{S}} x$ converges weakly for every $x \in \mathfrak{X}$ and some/every strong right \mathcal{S} -ergodic net $(A_{\alpha}^{\mathcal{S}})_{\alpha \in \Lambda}$,
- (3) $A_{\alpha}^{\mathcal{S}} x$ converges weakly for every $x \in \mathfrak{X}$ and some/every weak right \mathcal{S} -ergodic net $(A_{\alpha}^{\mathcal{S}})_{\alpha \in \Lambda}$,
- (4) $A_{\alpha}^{\mathcal{S}} x$ has a weak cluster point for every $x \in \mathfrak{X}$ and some/every weak right \mathcal{S} -ergodic net $(A_{\alpha}^{\mathcal{S}})_{\alpha \in \Lambda}$,
- (5) $\overline{\text{co}}\mathcal{S}x \cap \text{Fix}(\mathcal{S}) \neq \emptyset$ for all $x \in \mathfrak{X}$,
- (6) there exists operator $P \in \overline{\text{co}}\mathcal{S}$ such that $S_g P = P S_g = P^2 = P$ for every $g \in G$,
- (7) $\text{Fix}(\mathcal{S})$ separates $\text{Fix}(\mathcal{S}')$,
- (8) $\mathfrak{X} = \text{Fix}(\mathcal{S}) \oplus \overline{\text{span}} \text{rg}(I - \mathcal{S})$.

The strong (weak) limit $\lim_{\alpha} A_{\alpha}^{\mathcal{S}} x$ is equal for every strong (weak) right \mathcal{S} -ergodic net $(A_{\alpha}^{\mathcal{S}})_{\alpha \in \Lambda}$ and is also equal to Px . Operator P is a \mathcal{S} -absorbing projection onto $\text{Fix}(\mathcal{S})$ along $\overline{\text{span}} \text{rg}(I - \mathcal{S})$.

We will make use of the following simple facts.

Fact 4.1 Let \mathcal{S} be a bounded representation of semitopological semigroup G . If \mathcal{S} is mean ergodic then $\text{Fix}(\mathcal{S}')$ separates $\text{Fix}(\mathcal{S})$ (i.e. for every $0 \neq x \in \text{Fix}(\mathcal{S})$ there exists $x^* \in \text{Fix}(\mathcal{S}')$ with $\langle x, x^* \rangle \neq 0$).

Proof: Let $x \in \text{Fix}(\mathcal{S})$ be nonzero. There exists $x^* \in \mathfrak{X}^*$ with $\langle x, x^* \rangle \neq 0$ since the dual space of the normed space always separates the space. For the absorbing projection P we have $Px \in \overline{\text{co}}\mathcal{S}x = \{x\}$, hence

$$\langle x, P'x^* \rangle = \langle Px, x^* \rangle = \langle x, x^* \rangle \neq 0.$$

It remains to show that $P'x^* \in \text{Fix}(\mathcal{S}')$. It follows from

$$\langle y, S'_g P'x^* \rangle = \langle PS_g y, x^* \rangle = \langle Py, x^* \rangle = \langle y, P'x^* \rangle \text{ for every } y \in \mathfrak{X},$$

since $S_g P = P$ for every $g \in G$. \square

Fact 4.2 *Let V be a normed space and $E \subset V$, $F \subset V^*$ be linear subspaces. If E separates F then $\dim F \leq \dim E$ (i.e. if $\dim E = k < \infty$ then $\dim F \leq k$). Analogously, if F separates E then $\dim E \leq \dim F$.*

Proof: Assume that $\dim E = k < \infty$, $E = \text{span}\{x_1, \dots, x_k\}$ and F separates E . Let us define the linear map $\Lambda : F \rightarrow \mathbb{C}^k$ by

$$\Lambda x^* = (\langle x_1, x^* \rangle, \langle x_2, x^* \rangle, \dots, \langle x_k, x^* \rangle), \quad x^* \in F.$$

Note that Λ is injective. Indeed, $\Lambda x^* = 0$ implies $\langle x, x^* \rangle = 0$ for every $x \in E$, but E separates F , so $x^* = 0$. This shows that $\dim F \leq k$. If now F separates E and $F = \text{span}\{x_1^*, \dots, x_k^*\}$, we define $\Lambda : E \rightarrow \mathbb{C}^k$ by

$$\Lambda x = (\langle x, x_1^* \rangle, \langle x, x_2^* \rangle, \dots, \langle x, x_k^* \rangle), \quad x \in E.$$

Again, Λ is injective, so $\dim E \leq k$. \square

4.2. Amenable semigroups

Let X be a topological space. Denote by $C_b(X)$ the normed space of all bounded continuous complex-valued functions on X with the supremum norm.

Definition 4.5 Let G be a semitopological semigroup. We say that $m \in C_b(G)^*$ is a **mean** if $\langle \mathbb{1}, m \rangle = \|m\| = 1$. We say that a mean m is **right (left) invariant** if it satisfies

$$m(R_g f) = m(f) \quad \left(m(L_g f) = m(f) \right),$$

where for $g \in G$ we have $R_g, L_g : C_b(G) \rightarrow C_b(G)$, $R_g(f)(h) := f(hg)$, $L_g(f)(h) := f(hg)$. A mean m is called **invariant** if it's both left and right invariant. The semigroup G is called **right (left) amenable** if there exists right (left) invariant mean on G . G is called **amenable** if there exists mean m on G which is both left and right invariant.

Example 4.4 \mathbb{N} with discrete topology is a amenable semitopological semigroup.

We are interested in the amenability of the semigroup, since bounded representation of amenable semigroup admits ergodic net.

Theorem 4.4 ([Schreiber13, prop. 1.3])

Let \mathcal{S} be a bounded representation of G on \mathfrak{X} . If G is (right) amenable then there exists weak (right) \mathcal{S} -ergodic net.

Using Theorem 4.1 we can strengthen the previous result:

Corollary 4.1

Let \mathcal{S} be a bounded representation of G on \mathfrak{X} . If G is (right) amenable then there exists strong (right) \mathcal{S} -ergodic net.

Corollary 4.2 ([Schreiber13, cor. 1.9])

Let \mathcal{S} be a bounded representation of amenable semigroup G on \mathfrak{X} . If for every $x \in \mathfrak{X}$ the set Sx is relatively weakly compact then \mathcal{S} is mean ergodic.

Proof: By Krein-Smulian Theorem (1.10) we have that $\overline{\text{co}}Sx$ is weakly compact (note that norm and weak closure of convex set are equal). Let $(A_\alpha^S)_{\alpha \in \Lambda}$ be a weak ergodic net. Since $A_\alpha^S \in \overline{\text{co}}^{s.o.t.}(\mathcal{S})$, we have $A_\alpha^S x \in \overline{\text{co}}Sx$ for each $x \in \mathfrak{X}$, hence net $A_\alpha^S x$ has a weak cluster point for each $x \in \mathfrak{X}$. By Theorem 4.3 we have that \mathcal{S} is mean ergodic.

Using the fact that closed unit ball in a reflexive Banach space is weakly compact we obtain the following

Corollary 4.3

A bounded representation \mathcal{S} of amenable semigroup G on a reflexive Banach space \mathfrak{X} is mean ergodic. In particular every power bounded sequence $(T_n)_{n \in \mathbb{N}}$ of operators in $L(\mathfrak{X})$ with $T_{n+m} = T_m T_n$ for $n, m \in \mathbb{N}$ on reflexive Banach space is mean ergodic.

Note that von Neumann's Ergodic Theorem (Theorem 2.5) is a special case of the previous corollary, since Hilbert space is always reflexive.

4.3. Markov operators on $C(X)$

We will introduce now class of operators for which we will establish semigroup generalization of topological Wiener-Wintner theorem.

Definition 4.6 Let X be a compact Hausdorff topological space. Operator $S \in L(C(X))$ is called a **Markov operator** if $S\mathbb{1} = \mathbb{1}$ and S is positive, i.e. $Sf \geq 0$ for $f \geq 0$. $\mathbb{1} \in C(X)$ stands for a function with $\mathbb{1}(x) = 1$ for all $x \in X$. We will say that bounded representation \mathcal{S} of the semitopological semigroup G on $C(X)$ is **Markovian** if it consists of Markov operators.

Example 4.5 Let (X, T) be a topological dynamical system. The Koopman operator $U_T : C(X) \rightarrow C(X)$ is a Markov operator, since for $f \geq 0$ there is $f \circ T \geq 0$ and $\mathbb{1} \circ T(x) = \mathbb{1}(Tx) = 1$, so $U_T \mathbb{1} = \mathbb{1}$.

Remark Let S be a Markov operator on $C(X)$. Its dual operator $S' : \mathcal{M}(X) \rightarrow \mathcal{M}(X)$ is also positive and for all $\mu \in \mathcal{M}(X)$ we have $S'\mu(X) = \mu(X)$. Indeed, for $\mathcal{M}(X) \ni \mu \geq 0$ and $C(X) \ni f \geq 0$ we have

$$\int_X f dS'\mu = \langle f, S'\mu \rangle = \langle Sf, \mu \rangle = \int_X Sf d\mu \geq 0,$$

so $S'\mu \geq 0$. There also is

$$S'\mu(X) = \int_X \mathbb{1} dS'\mu = \langle \mathbb{1}, S'\mu \rangle = \langle S\mathbb{1}, \mu \rangle = \langle \mathbb{1}, \mu \rangle = \mu(X).$$

Note that if μ is a probability measure on X then $S'\mu$ is also a probability measure. This allows one to think about S' as an operator describing evolution of probabilities on space X . In particular, if δ_x is a unit mass measure in a point $x \in X$ then $S'\delta_x$ is a probability which describes stochastic evolution of the point x .

We will give now some lattice properties of Markov operator and its dual operator. For the definition of the modulus of complex measure recall Definition 1.4. By $\mu \leq \nu$, $\mu, \nu \in \mathcal{M}(X)$ we mean that measures μ and ν are real-valued and $\nu - \mu$ is nonnegative ($\nu(A) \geq \mu(A)$ for every $A \in \mathcal{B}(X)$).

Lemma 4.1

Let X be a compact Hausdorff topological space. Let $S : C(X) \rightarrow C(X)$ be a Markov operator. The following holds:

- (1) if $f \in C(X)$ is real-valued, then Sf is also real-valued,
- (2) for every $f \in C(X)$ there is $S(\operatorname{Re} f) = \operatorname{Re}(Sf)$ and $S(\operatorname{Im} f) = \operatorname{Im}(Sf)$,
- (3) $|Sf| \leq S|f|$ for every $f \in C(X)$
- (4) $|S'\mu| \leq S'|\mu|$ for every $\mu \in \mathcal{M}(X)$,
- (5) $\|S\| = 1$ and $\|S'\| = 1$,
- (6) for a given $0 \leq \mu \in \mathcal{M}(X)$, if $S'\mu \geq \mu$ then $S'\mu = \mu$.

Proof: (1) We have $f = f^+ - f^-$, $f^+, f^- \geq 0$ and $Sf = Sf^+ - Sf^-$. Since S is positive, we have that $Sf^+, Sf^- \geq 0$, so Sf is also real-valued.

(2) We have $f = \operatorname{Re} f + i\operatorname{Im} f$ and $Sf = S(\operatorname{Re} f) + S(i\operatorname{Im} f) = S(\operatorname{Re} f) + iS(\operatorname{Im} f)$. From the previous point we know that $S(\operatorname{Re} f)$ and $S(\operatorname{Im} f)$ are real-valued. Since of course $Sf = \operatorname{Re}(Sf) + i\operatorname{Im}(Sf)$ and splitting to the real and imaginary part is unique, there must be $\operatorname{Re}(Sf) = S(\operatorname{Re} f)$ and $\operatorname{Im}(Sf) = S(\operatorname{Im} f)$.

(3) Note first that for $f \in C(X)$ we have $|f|(x) = \sup_{\lambda \in \mathbb{T}} \operatorname{Re}(\lambda f(x))$. For every $\lambda \in \mathbb{T}$ there is of course $\operatorname{Re}(\lambda f) \leq |f|$, so also $S\operatorname{Re}(\lambda f) = \operatorname{Re}(\lambda Sf) \leq S|f|$. Passing to the supremum yields $|Sf| = \sup_{\lambda \in \mathbb{T}} \operatorname{Re}(\lambda Sf) \leq S|f|$.

(4) For $\mu \in \mathcal{M}(X)$ and $0 \leq f \in C(X)$ there is by Fact 1.2 and the previous point

$$\begin{aligned} \langle f, |S'\mu| \rangle &= \sup\{|\langle h, S'\mu \rangle| : h \in C(X), |h| \leq f\} = \sup\{|\langle Sh, \mu \rangle| : h \in C(X), |h| \leq f\} \leq \\ &\sup\{|\langle Sh, |\mu| \rangle| : h \in C(X), |h| \leq f\} \leq \sup\{\langle S|h|, |\mu| \rangle : h \in C(X), |h| \leq f\} \leq \\ &\leq \langle Sf, |\mu| \rangle = \langle f, S'|\mu| \rangle. \end{aligned}$$

(5) For $f \in C(X)$ there is $\|Sf\| = \|S|f|\| \leq \|S\| \|f\| \leq \|S\| \|f\| \cdot 1 = \|f\| \|S\| = \|f\|$, so $\|S\| \leq 1$. $S\mathbf{1} = \mathbf{1}$ yields now that $\|S\| = 1$. Since $\|S\| = \|S'\|$, there is also $\|S'\| = 1$.

(6) For every $0 \leq f \in C(X)$ there is

$$\begin{aligned} 0 \leq \langle f, S'\mu - \mu \rangle &\leq \langle \|f\|_{\sup} \cdot \mathbf{1}, S'\mu - \mu \rangle = \langle \|f\|_{\sup} \cdot \mathbf{1}, S'\mu \rangle - \langle \|f\|_{\sup} \cdot \mathbf{1}, \mu \rangle = \\ &= \|f\|_{\sup} (\langle S\mathbf{1}, \mu \rangle - \langle \mathbf{1}, \mu \rangle) = 0, \end{aligned}$$

hence $S'\mu = \mu$. \square

Now we would like to investigate subsets of X invariant for the Markov operators. Note that we have to give new definition of invariance, since we don't necessarily consider deterministic transformation on X .

Definition 4.7 Let X be a compact Hausdorff topological space and let $S : C(X) \rightarrow C(X)$ be a Markov operator. We say that subset $F \subset X$ is **invariant** (for S) if F is nonempty, closed and for every $x \in F$ there is $\text{supp}(S'\delta_x) \subset F$. If we consider semigroup \mathcal{S} of Markov operators, then F is called invariant if it is invariant for every operator $S \in \mathcal{S}$. Invariant set F is called **minimal** if there exists no invariant proper subset of F .

Properties of invariant sets for a single Markov operator can be found in [Sine]. We give analogous properties of sets invariant for semigroup of Markov operators.

Proposition 4.1

Let X be a compact Hausdorff topological space and let $\mathcal{S} = \{S_g : g \in G\}$ be a bounded, Markovian representation of semitopological semigroup G on $C(X)$. Then closed and nonempty set $F \subset X$ is invariant if and only if $S_g f$ vanishes on F for every $g \in G$ and every $f \in C(X)$ vanishing on F .

Proof: Assume that F is invariant and $C(X) \ni f = 0$ on F . For every $g \in G$ and $x \in F$ we have

$$S_g f(x) = \langle S_g f, \delta_x \rangle = \langle f, S'_g \delta_x \rangle = \int_X f dS'_g \delta_x = \int_{\text{supp}(S'_g \delta_x)} f dS'_g \delta_x = 0,$$

since $\text{supp}(S'_g \delta_x) \subset F$ and f vanishes on F . Conversely, assume that F is not invariant, i.e. there exists $x \in F$ and $g \in G$ with $y \in \text{supp}(S'_g \delta_x) \setminus F$ for some $y \in X$. Take continuous function $f : X \rightarrow [0, 1]$ with $f(y) = 1$ and $f = 0$ on F (it exists due to the Urysohn's lemma (Theorem 1.11)). We have then

$$S_g f(x) = \int_X S_g f d\delta_x = \int_X f dS'_g \delta_x > 0,$$

since $f(y) = 1$ and $y \in \text{supp}(S'_g \delta_x)$ (there must be $f > \frac{1}{2}$ on some neighborhood U of y and since y is in the support of $S'_g \delta_x$, there must be $S'_g \delta_x(U) > 0$. Both the function f and the measure $S'_g \delta_x$ are nonnegative, hence the integral must be strictly positive). There is $x \in F$, so $S_g f$ doesn't vanish on F . \square

Remark Let F be an invariant set. For every $g \in G$ and $f, h \in C(X)$ we have $S_g f = S_g h$ if $f = h$ on F . This allows us to consider restricted semigroup $\mathcal{S}|_F := \{S_g|_F : g \in G\}$ of Markov operators on $C(F)$, where for $f \in C(F)$ we have $S_g|_F(f) := (S_g h)|_F$ for any $h \in C(X)$ with $f = h$ on F . The previous proposition assures that this definition is correct. $\mathcal{S}|_F$ is a bounded representation of G on $C(F)$. If \mathcal{S} admits (right) ergodic net, then also $\mathcal{S}|_F$ admits (right) ergodic net (it's enough to restrict \mathcal{S} -ergodic net to F). If \mathcal{S} is in addition mean ergodic, then $\mathcal{S}|_F$ is also mean ergodic. To see that take $0 \neq \mu \in \text{Fix}(\mathcal{S}'|_F)$ and consider measure $\nu(A) := \mu(A \cap F)$, $A \in \mathcal{B}(X)$. For $f \in C(X)$ there is

$$\langle f, S'_g \nu \rangle = \langle S_g f, \nu \rangle = \int_X S_g f d\nu = \int_F S_g f d\nu = \langle (S_g f)|_F, \mu \rangle = \langle S_g|_F(f|_F), \mu \rangle = \langle f|_F, \mu \rangle = \langle f, \nu \rangle,$$

so $\nu \in \text{Fix}(\mathcal{S}')$. There exists $f \in \text{Fix}(\mathcal{S})$ with $\langle f, \nu \rangle \neq 0$, but again $S_g|_F(f|_F) = (S_g f)|_F = f|_F$, so $f|_F \in \text{Fix}(\mathcal{S}|_F)$ and

$$\langle f|_F, \mu \rangle = \int_F f|_F d\mu = \int_X f d\nu = \langle f, \nu \rangle \neq 0,$$

hence $\text{Fix}(\mathcal{S}|_F)$ separates $\text{Fix}(\mathcal{S}'|_F)$ and $\mathcal{S}|_F$ is mean ergodic. Note also that if $K \subset F$ is invariant for $\mathcal{S}|_F$ then K is invariant also for \mathcal{S} . Indeed, assume that $f \in C(X)$ vanishes on K . We have that $f|_F \in C(F)$ vanishes on K , so $S_g|_F(f|_F)$ vanishes on K . On the other hand $S_g|_F(f|_F) = (S_g f)|_F$, so in fact $S_g f$ vanishes on K . Proposition 4.1 implies that K is invariant for \mathcal{S} .

Lemma 4.2

Let X be a compact Hausdorff topological space and let $\mathcal{S} = \{S_g : g \in G\}$ be a bounded, Markovian representation of semitopological semigroup G on $C(X)$ which admits strong right ergodic net. The following conditions holds:

- (1) if F_1, F_2 are invariant sets, then $F_1 \cap F_2$ is invariant if nonempty
- (2) if F is invariant and for a probability measure $\mu \in \mathcal{M}^1(X)$ there is $\text{supp}(\mu) \subset F$ then also $\text{supp}(S'_g \mu) \subset F$ for every $g \in G$,
- (3) $\mathbb{P}_{\mathcal{S}}$ is a convex and weak* compact subset of $\mathcal{M}(X)$,
- (4) $\text{supp}(\mu)$ is invariant for every $\mu \in \mathbb{P}_{\mathcal{S}}$,
- (5) for every real-valued $f \in \text{Fix}(\mathcal{S})$ sets $M(f) := \{x \in X : f(x) = \sup_{y \in X} f(y)\}$ and $m(f) := \{x \in X : f(x) = \inf_{y \in X} f(y)\}$ are invariant,
- (6) for every real-valued $f \in \text{Fix}(\mathcal{S})$ and $\mu \in \mathbb{P}_{\mathcal{S}}$ sets $\{f \leq r\} \cap \text{supp}(\mu)$ and $\{f \geq r\} \cap \text{supp}(\mu)$ are invariant if nonempty,
- (7) if \mathcal{S} is mean ergodic, then for every invariant $F \subset X$ there exists $\mu \in \mathbb{P}_{\mathcal{S}}$ with $\text{supp}(\mu) \subset F$,
- (8) if \mathcal{S} is mean ergodic, then for minimal set F every $f \in \text{Fix}(\mathcal{S})$ is constant on F ,
- (9) if \mathcal{S} is mean ergodic, then for $\mu \in \mathbb{P}_{\mathcal{S}}$ its support $\text{supp}(\mu)$ is minimal if and only if every $f \in \text{Fix}(\mathcal{S})$ is constant on $\text{supp}(\mu)$,
- (10) if \mathcal{S} is mean ergodic, then $\text{supp}(\mu)$ is minimal for every $\mu \in \text{Ext}(\mathbb{P}_{\mathcal{S}})$.

Proof: (1) $F_1 \cap F_2$ is closed and for $x \in F_1 \cap F_2$ we have $\text{supp}(S'_g \delta_x) \subset F_1$ by the invariance of F_1 and $\text{supp}(S'_g \delta_x) \subset F_2$ by the invariance of F_2 , hence $\text{supp}(S'_g \delta_x) \subset F_1 \cap F_2$ for every $g \in G$.

(2) By Fact 2.4 it suffices to prove that $S'_g \mu(F) = 1$ for every $g \in G$. Let us fix $\varepsilon > 0$. By regularity of the measure there exists open set U with $F \subset U$ and $S'_g \mu(U \setminus F) < \varepsilon$. By Urysohn's lemma (Theorem 1.11) there exists $f \in C(X)$ with $\|f\|_{\text{sup}} = 1$, $f = 1$ on F and $f = 0$ on U' . There is

$$S'_g \mu(F) = \int_F 1 dS'_g \mu = \int_F f dS'_g \mu = \int_X f dS'_g \mu - \int_{U \setminus F} f dS'_g \mu = \int_X S_g f d\mu - \int_{U \setminus F} f dS'_g \mu = (*)$$

By Proposition 4.1 we have $S_g f = S_g \mathbb{1} = \mathbb{1}$ on F and since $\text{supp}(\mu) \subset F$ there is further

$$(*) = \int_F S_g f d\mu - \int_{U \setminus F} f dS'_g \mu = \int_F \mathbb{1} d\mu - \int_{U \setminus F} f dS'_g \mu = \mu(F) - \int_{U \setminus F} f dS'_g \mu = 1 - \int_{U \setminus F} f dS'_g \mu.$$

We also have

$$\left| \int_{U \setminus F} dS'_g \mu \right| \leq \int_{U \setminus F} |f| dS'_g \mu \leq S'_g \mu(U \setminus F) \|f\|_{\text{sup}} < \varepsilon.$$

We finally obtain $|S'_g \mu(F) - 1| < \varepsilon$ and since ε was arbitrary, we have $S'_g \mu(F) = 1$.

- (3) For a probability measure $\mu \in \mathcal{M}^1(X)$ there is $\|\mu\| = \mu(X) = 1$, so $\mathbb{P}_{\mathcal{S}} \subset \overline{B}(0, 1)$. By the Banach-Alaoglu theorem $\overline{B}(0, 1)$ is weak* compact, so it suffices to show that $\mathbb{P}_{\mathcal{S}}$ is weak* closed. Let $(\mu_\alpha)_{\alpha \in \Lambda}$ be a net of measures with $\mu_\alpha \in \mathbb{P}_{\mathcal{S}}$ for every $\alpha \in \Lambda$ and suppose that $\mu_\alpha \xrightarrow{*w} \mu \in \mathcal{M}(X)$. For $0 \leq f \in C(X)$ there is

$$\langle f, \mu \rangle = \lim_{\alpha} \langle f, \mu_\alpha \rangle \geq 0,$$

so $\mu \geq 0$. There also is

$$\mu(X) = \langle \mathbb{1}, \mu \rangle = \lim_{\alpha} \langle \mathbb{1}, \mu_\alpha \rangle = \lim_{\alpha} 1 = 1,$$

so μ is a probability measure. For any $f \in C(X)$ and $g \in G$ there is

$$\langle f, S'_g \mu \rangle = \langle S_g f, \mu \rangle = \lim_{\alpha} \langle S_g f, \mu_\alpha \rangle = \lim_{\alpha} \langle f, \mu_\alpha \rangle = \langle f, \mu \rangle,$$

so μ is invariant, hence $\mu \in \mathbb{P}_{\mathcal{S}}$.

- (4) Let $0 \leq f \in C(X)$ vanish on $\text{supp}(\mu)$. There is

$$\int_{\text{supp}(\mu)} S_g f d\mu = \int_X S_g f d\mu = \langle S_g f, \mu \rangle = \langle f, S'_g \mu \rangle = \langle f, \mu \rangle = \int_X f d\mu = \int_{\text{supp}(\mu)} f d\mu = 0.$$

Since $S_g f$ and μ are nonnegative, there must be $S_g f = 0$ on $\text{supp}(\mu)$. If f is real-valued and vanishes on $\text{supp}(\mu)$ then both f^+ and f^- must vanish on $\text{supp}(\mu)$, hence $Sf = Sf^+ - Sf^-$ vanish on $\text{supp}(\mu)$. Finally, if f is complex-valued, then both $\text{Re} f$ and $\text{Im} f$ must vanish on $\text{supp}(\mu)$ and the same argument works. Note that $\text{supp}(\mu)$ is closed and $\text{supp}(\mu) \neq \emptyset$, since $\mu(\text{supp}(\mu)) = 1$ (Fact 2.4).

- (5) We will prove only that $m(f)$ is invariant, proof for $M(f)$ is analogous. It is clear that $m(f)$ is closed and nonempty. Set $r := \inf_{y \in X} f(y)$ and take $x \in m(f)$. For every $g \in G$ we have

$$r = f(x) = S_g f(x) = \int_X f dS'_g \delta_x,$$

but $f \geq r$ on X , so there must be $f = r$ on $\text{supp}(S'_g \delta_x)$. This proves that $\text{supp}(S'_g \delta_x) \subset m(f)$.

- (6) We will prove the fact only for $\{f \leq r\} \cap \text{supp}(\mu)$. It suffices to show that $\{f \leq r\} \cap \text{supp}(\mu)$ is invariant for $\mathcal{S}|_{\text{supp}(\mu)}$. Consider $h \in C(X)$ given as $h = \max\{f, r\}$. We have $m(h) \cap \text{supp}(\mu) = \{f \leq r\} \cap \text{supp}(\mu)$. There is $h \geq f$ and $h \geq r \cdot \mathbf{1}$, so $S_g h \geq S_g f = f$ and $S_g h \geq S_g r \cdot \mathbf{1} = r \cdot \mathbf{1}$, thus $S_g h \geq \max\{f, r\} = h$. On the other hand

$$\int_X S_g h d\mu = \int_X h dS'_g \mu = \int_X h d\mu,$$

hence $S_g h = h$ on $\text{supp}(\mu)$. Since h is invariant on supp and the previous point gives invariance of $m(h) \cap \text{supp}(\mu)$.

- (7) We will begin with proving that $\mathbb{P}_{\mathcal{S}} \neq \emptyset$. Let P be a \mathcal{S} -absorbing projection with $P \in \overline{co}\mathcal{S}$. Note that P is a Markov operator. Indeed, let $\{A_\alpha\}_{\alpha \in \Lambda}$ be a net of operators from $co\mathcal{S}$ with $A_\alpha \xrightarrow{s.o.t.} P$. Convex combination of Markov operators is Markovian, hence every A_α , $\alpha \in \Lambda$ is a Markov operator. We have $P\mathbf{1} = \lim_{\alpha} A_\alpha \mathbf{1} = \lim_{\alpha} \mathbf{1} = \mathbf{1}$ and for $0 \leq f \in C(X)$ there is $Pf = \lim_{\alpha} A_\alpha f \geq 0$, so P is a Markov operator. Now the adjoint operator P' takes probability measures into probability measures, so $P'\delta_x$ is a probability measure. $P'\delta_x$ is also invariant, since

$$\langle f, S'_g P' \delta_x \rangle = \langle P S_g f, \delta_x \rangle = \langle P f, \delta_x \rangle = \langle f, P' \delta_x \rangle,$$

for every $f \in C(X)$, $g \in G$. Let F be an invariant subset of X . $\mathcal{S}|_F$ is mean ergodic, since \mathcal{S} is mean ergodic, so there exists $\mu \in \mathbb{P}_{\mathcal{S}|_F}$. We can extend the measure $\mu \in \mathcal{M}(F)$ to the measure $\nu \in \mathcal{M}(X)$ by $\nu(A) := \mu(A \cap F)$ for $A \in \mathcal{B}(X)$. We already showed that ν is \mathcal{S} -invariant (remark after Proposition 4.1) and obviously ν is a probability measure, so $\nu \in \mathbb{P}_{\mathcal{S}}$ and there is $\text{supp}(\nu) \subset F$.

- (8) Suppose that F is minimal. There exists $\mu \in \mathbb{P}_{\mathcal{S}}$ with $\text{supp}(\mu) \subset F$, but there must actually be $\text{supp}(\mu) = F$, since F is minimal. If there would be $f \in \text{Fix}(\mathcal{S})$ non constant on F , then $\text{Re}(f)$ and $\text{Im}(f)$ are both invariant and one of them must be non constant, so we can assume that f is real valued. Then we have $\emptyset \neq \{f \leq r\} \cap F \subsetneq F$ for some $r \in \mathbb{R}$ and by (6) $\{f \leq r\} \cap F = \{f \leq r\} \cap \text{supp}(\mu)$ is invariant, what gives a contradiction with the minimality of F .
- (9) If $\text{supp}(\mu)$ is minimal, then every $f \in \text{Fix}(\mathcal{S})$ is constant on $\text{supp}(\mu)$ by (8). Suppose that $\text{supp}(\mu)$ is not minimal. There exists invariant $F \subsetneq \text{supp}(\mu)$. By (7) there exists $\nu \in \mathbb{P}_{\mathcal{S}}$ with $\text{supp}(\nu) \subset F \subsetneq \text{supp}(\mu)$, so $\mu \neq \nu$. Since \mathcal{S} is mean ergodic, there exists $f \in \text{Fix}(\mathcal{S})$ with $\langle f, \mu \rangle \neq \langle f, \nu \rangle$. f cannot be constant on $\text{supp}(\mu)$, since then there would be

$$\langle f, \mu \rangle = \int_X f d\mu = \int_{\text{supp}(\mu)} f d\mu = c = \int_{\text{supp}(\mu)} f d\nu = \int_X f d\nu = \langle f, \nu \rangle,$$

for some constant $c \in \mathbb{C}$.

- (10) Suppose that $\text{supp}(\mu)$ is not minimal for some $\mu \in \text{Ext}(\mathbb{P}_{\mathcal{S}})$. By (9) there exists $f \in \text{Fix}(\mathcal{S})$ which is non constant on $\text{supp}(\mu)$. We can suppose that f is real-valued. There exists $r \in \mathbb{R}$ with $0 < \mu(\{f \leq r\}) < 1$. Indeed, since f is non constant on $\text{supp}(\mu)$, there exist $x, y \in \text{supp}(\mu)$ with $f(x) < r < f(y)$ for some $r \in \mathbb{R}$. There is $0 < \mu(\{f < r\}) \leq \mu(\{f \leq r\}) = 1 - \mu(\{f > r\}) < 1$, since $\{f < r\}$ and $\{f > r\}$ are open sets containing x and y

respectively and these points are taken from the support of μ . Denote $F := \{f \leq r\} \cap \text{supp}(\mu)$ and $V := \{f > r\} \cap \text{supp}(\mu)$. F is invariant (by (6)) and there is $\mu(F) = \mu(\{f \leq r\}) \in (0, 1)$ and $\mu(V) = \mu(\{f > r\}) \in (0, 1)$. Consider now non-negative probability measures $\nu_1, \nu_2 \in \mathcal{M}(X)$ given by

$$\mu \mathbb{1}_F(A) := \mu(A \cap F) \quad \text{and} \quad \mu \mathbb{1}_V := \mu(A \cap V), \quad A \in \mathcal{B}(X).$$

We will prove that $\mu \mathbb{1}_F$ and $\mu \mathbb{1}_V$ are invariant. There obviously is $0 \leq \mu \mathbb{1}_F \leq \mu$ and $0 \leq \mu \mathbb{1}_V \leq \mu$, hence there is $0 \leq S'_g(\mu \mathbb{1}_F) \leq S'_g \mu = \mu$ and $0 \leq S'_g(\mu \mathbb{1}_V) \leq S'_g \mu = \mu$ for every $g \in G$. There is $\text{supp}(\mu \mathbb{1}_F) \subset F$, hence 2 gives that $\text{supp}(S'_g(\mu \mathbb{1}_F)) \subset F$, hence

$$S'_g(\mu \mathbb{1}_F)(A) = S'_g(\mu \mathbb{1}_F)(A \cap F) \leq \mu(A \cap F) = \mu \mathbb{1}_F(A) \quad \text{for every } A \in \mathcal{B}(X),$$

so in fact $S'_g(\mu \mathbb{1}_F) \leq \mu \mathbb{1}_F$. Fact that $S'_g(\mu \mathbb{1}_V) \leq \mu \mathbb{1}_V$ requires different argument. We have

$$V = \{f > r\} \cap \text{supp}(\mu) = \bigcup_{n=1}^{\infty} \{f \geq r + \frac{1}{n}\} \cap \text{supp}(\mu).$$

Denote $K_N := \bigcup_{n=1}^N \{f \geq r + \frac{1}{n}\} \cap \text{supp}(\mu)$. There is $0 < \mu(V) = \lim_{N \rightarrow \infty} \mu(K_N)$, so $K_N \neq \emptyset$ and $\mu(K_N) > 0$ for large enough $N \in \mathbb{N}$. Note that each K_N is invariant (if nonempty), since is a finite sum of invariant sets. In particular there is $\text{supp}(S'_g \mu \mathbb{1}_{K_N}) \subset \text{supp}(\mu \mathbb{1}_{K_N}) \subset K_N$ (by (2)), hence $S'_g(\mu \mathbb{1}_{K_N})(K_N) = \mu \mathbb{1}_{K_N}(K_N) = \mu(K_N)$. We have now (for large enough N)

$$S'_g(\mu \mathbb{1}_V)(V) \geq S'_g(\mu \mathbb{1}_{K_N})(K_N) = \mu(K_N),$$

hence passing to limit yields

$$\mu(V) = \mu \mathbb{1}_V(V) = \mu \mathbb{1}_V(X) = S'_g(\mu \mathbb{1}_V)(X) \geq S'_g(\mu \mathbb{1}_V)(V) \geq \lim_{N \rightarrow \infty} \mu(K_N) = \mu(V).$$

We can now continue with

$$S'_g(\mu \mathbb{1}_V)(A) = S'_g(\mu \mathbb{1}_V)(A \cap V) \leq \mu(A \cap V) = \mu \mathbb{1}_V(A) \quad \text{for every } A \in \mathcal{B}(X),$$

so $S'_g \mu \mathbb{1}_V \leq \mu \mathbb{1}_V$. Suppose now that $S'_g(\mu \mathbb{1}_F) < \mu \mathbb{1}_F$ or $S'_g(\mu \mathbb{1}_V) < \mu \mathbb{1}_V$ for some $g \in G$. We would have then

$$\mu = S'_g \mu = S'_g(\mu \mathbb{1}_F + \mu \mathbb{1}_V) = S'_g(\mu \mathbb{1}_F) + S'_g(\mu \mathbb{1}_V) < \mu \mathbb{1}_F + \mu \mathbb{1}_V = \mu,$$

what gives a contradiction. We finally have

$$\frac{\mu \mathbb{1}_F}{\mu(F)}, \frac{\mu \mathbb{1}_V}{\mu(V)} \in \mathbb{P}_{\mathcal{S}} \quad \text{and} \quad \mu = \mu(F) \frac{\mu \mathbb{1}_F}{\mu(F)} + \mu(V) \frac{\mu \mathbb{1}_V}{\mu(V)},$$

hence μ cannot be in $\text{Ext}(\mathbb{P}_{\mathcal{S}})$. \square

For a given Markov operator S on $C(X)$ and invariant probability measure $\mu \in \mathbb{P}_{\mathcal{S}}$ we are interested in extending S to the linear contractive operator on $L^2(\mu)$. Note first, that for $f \in C(X)$ ($\subset L^1(\mu)$) we have

$$\|f\|_{L^1(\mu)} = \int_X |f| d\mu = \langle |f|, \mu \rangle = \langle |f|, S' \mu \rangle = \langle S|f|, \mu \rangle \geq \langle |Sf|, \mu \rangle = \|Sf\|_{L^1(\mu)},$$

hence S is a bounded linear operator on $C(X)$ seen as a subspace of $L^1(\mu)$. Since $C(X)$ is dense in $L^1(\mu)$ (in L^1 norm), we can uniquely extend S to a positive contraction $\hat{S} : L^1(\mu) \rightarrow L^1(\mu)$. Moreover, since $S\mathbf{1} = \mathbf{1}$, we have for $f \in L^\infty(\mu)$

$$\begin{aligned}\|\hat{S}f\|_{L^\infty(\mu)} &= \|\hat{S}f\|_{L^\infty(\mu)} \leq \|\hat{S}\|_{L^\infty(\mu)} \|f\|_{L^\infty(\mu)} = \|\hat{S}(\|f\|_{L^\infty(\mu)} \cdot \mathbf{1})\|_{L^\infty(\mu)} = \|S(\|f\|_{L^\infty(\mu)} \cdot \mathbf{1})\|_{L^\infty(\mu)} = \\ &= \|f\|_{L^\infty(\mu)} \|S\mathbf{1}\|_{L^\infty(\mu)} = \|f\|_{L^\infty(\mu)} \|\mathbf{1}\|_{L^\infty(\mu)} = \|f\|_{L^\infty},\end{aligned}$$

hence \hat{S} is also a $\|\cdot\|_{L^\infty(\mu)}$ -contraction (restricted to the $L^\infty(\mu)$). We can now make use of the following

Theorem 4.5 ([Eisner et al, thm 8.23])

Let (X, \mathcal{A}, μ) be a measure space and assume that $T : L^1(\mu) \rightarrow L^1(\mu)$ is a contraction and

$$\|Tf\|_{L^\infty(\mu)} \leq \|f\|_{L^\infty(\mu)} \text{ for all } f \in L^1(\mu) \cap L^\infty(\mu).$$

Then for every $f \in L^1(\mu) \cap L^p(\mu)$, $p \in (1, \infty)$ we have

$$\|Tf\|_{L^p(\mu)} \leq \|f\|_{L^p(\mu)}.$$

In our case (X, \mathcal{A}, μ) is a probability space, so $L^p(\mu) \subset L^1(\mu)$ for every $p \in (1, \infty)$. We can conclude that \hat{S} restricted to the $L^p(\mu)$ is a contraction in $L^p(\mu)$ norm. In particular, we can uniquely extend Markov operator S (on $C(X)$) to the positive contraction on $L^2(\mu)$ (by extending first to $L^1(\mu)$ and then restricting to $L^2(\mu)$). Note that we made use of the invariance of the measure μ .

In the case of invariant probability measure $\mu \in \mathbb{P}_S$ for bounded Markovian representation \mathcal{S} of semitopological semigroup G on $C(X)$ we can consider semigroup of positive contractions on $L^2(\mu)$. We denote it by $\mathcal{S}_{2,\mu} := \{S_{g,2} : g \in G\}$, where $S_{g,2}$ is the extension of S_g to $L^2(\mu)$. Note that $\mathcal{S}_{2,\mu}$ is a bounded representation of G on $L^2(\mu)$. Boundedness is obvious since since $S_{g,2}$ in $\mathcal{S}_{2,\mu}$ is a contraction, condition $S_{g,2}S_{h,2} = S_{hg,2}$ is satisfied since it is satisfied for the semigroup \mathcal{S} . In order to check the last condition, i.e. continuity in $L^2(\mu)$ of $G \ni g \mapsto S_{g,2}f$ for every $f \in L^2(\mu)$, let us fix $f \in L^2(\mu)$, $g \in G$ and $\varepsilon > 0$. $C(X)$ is dense in $L^2(\mu)$, so there exists $f_0 \in C(X)$ with $\|f - f_0\|_{L^2(\mu)} < \varepsilon$. Since \mathcal{S} is a bounded representation of G on $C(X)$, there exists an open neighbourhood U of g such that there is $\|S_g f_0 - S_h f_0\|_{sup} < \varepsilon$ for every $h \in U$. For $h \in U$ we have now

$$\begin{aligned}\|S_{g,2}f - S_{h,2}f\|_{L^2(\mu)} &\leq \|S_{g,2}f - S_{g,2}f_0\|_{L^2(\mu)} + \|S_{g,2}f_0 - S_{h,2}f_0\|_{L^2(\mu)} + \\ &+ \|S_{h,2}f_0 - S_{h,2}f\|_{L^2(\mu)} \leq \|S_{g,2}\| \cdot \|f - f_0\|_{L^2(\mu)} + \|S_g f_0 - S_h f_0\|_{L^2(\mu)} + \\ &+ \|S_{h,2}\| \cdot \|f - f_0\|_{L^2(\mu)} \leq 2\|f - f_0\|_{L^2(\mu)} + \|S_g f_0 - S_h f_0\|_{sup} < 3\varepsilon.\end{aligned}$$

Together with the semigroup $\mathcal{S}_{2,\mu}$ we will consider its Hermitian adjoint semigroup $\mathcal{S}_{2,\mu}^* := \{S_{g,2}^* : g \in G\}$. Note that if G is amenable, then by Corollary 4.3 we have that $\mathcal{S}_{2,\mu}$ is mean ergodic (even if \mathcal{S} is not mean ergodic), since $L^2(\mu)$ is a Hilbert space (hence reflexive Banach space).

There is a natural extension of the concept of unique ergodicity to the semigroup of operators on $C(X)$:

Definition 4.8 Let X be a compact Hausdorff topological space. Let \mathcal{S} be a bounded representation of semitopological semigroup G on the Banach space $C(X)$. We say that \mathcal{S} is **uniquely ergodic** if there exists a probability measure $\mu \in \mathcal{M}(X)$ with $\text{Fix}(\mathcal{S}') = \mathbb{C} \cdot \mu$.

Fact 4.3 ([Schreiber14, prop. 2.2]) *Let X be a compact Hausdorff topological space. Let \mathcal{S} be a bounded Markovian representation of right amenable semitopological semigroup G on the Banach space $C(X)$. \mathcal{S} is uniquely ergodic if and only if \mathcal{S} is mean ergodic and $\text{Fix}(\mathcal{S}) = \mathbb{C} \cdot \mathbf{1}$.*

Proof: " \Rightarrow " Assume that \mathcal{S} is uniquely ergodic with the unique measure μ . For every $c \cdot \mu \in \text{Fix}(\mathcal{S}')$, $c \in \mathbb{C} \setminus \{0\}$ there is $\langle \mathbf{1}, c \cdot \mu \rangle = c \neq 0$ and $\mathbf{1} \in \text{Fix}(\mathcal{S})$, so $\text{Fix}(\mathcal{S})$ separates $\text{Fix}(\mathcal{S}')$, hence \mathcal{S} is mean ergodic by the Abstract mean ergodic theorem (Theorem 4.2). It follows from Fact 4.1 that $\text{Fix}(\mathcal{S}')$ separates $\text{Fix}(\mathcal{S})$. Fact 4.2 gives now $\dim \text{Fix}(\mathcal{S}) \leq \dim \text{Fix}(\mathcal{S}') = 1$, so $\mathbb{C} \cdot \mathbf{1} \subset \text{Fix}(\mathcal{S})$ yields $\text{Fix}(\mathcal{S}) = \mathbb{C} \cdot \mathbf{1}$.

" \Leftarrow " Assume now that \mathcal{S} is mean ergodic and $\text{Fix}(\mathcal{S}) = \mathbb{C} \cdot \mathbf{1}$. By Lemma 4.2.7 there exists invariant probability measure $\mu \in \mathbb{P}_{\mathcal{S}}$, so $\mathbb{C} \cdot \mu \subset \text{Fix}(\mathcal{S}')$. Mean ergodicity of \mathcal{S} yields that $\text{Fix}(\mathcal{S})$ separates $\text{Fix}(\mathcal{S}')$ and by Fact 4.2 there is $\dim \text{Fix}(\mathcal{S}') \leq \dim \text{Fix}(\mathcal{S}) = 1$, so in fact $\mathbb{C} \cdot \mu = \text{Fix}(\mathcal{S}')$. \square

Chapter 5

Wiener-Wintner theorems for semigroups of Markov operators on $C(X)$

In this chapter we will establish generalization the of topological Wiener-Wintner theorem to the case of Markovian bounded representation of amenable semigroup. Our starting point is the generalization for uniquely ergodic semigroup due to M. Schreiber ([Schreiber14]). We generalize it further without the assumption of unique ergodicity, simultaneously giving more elementary proof.

5.1. Characters and distorted semigroups

We begin with introducing the concept of the semigroup character.

Definition 5.1 Let G be a semitopological semigroup. A continuous map $\chi : G \rightarrow \mathbb{T}$ is called a **character of G** if it satisfies $\chi(gh) = \chi(g)\chi(h)$ for all $g, h \in G$. We denote the set of all characters of the semigroup G by \widehat{G} .

For a given semitopological semigroup G , its bounded representation \mathcal{S} of G on the Banach space \mathfrak{X} and a character χ of G we will be interested in the mean ergodicity of the **distorted semigroup** ${}_{\chi}\mathcal{S} := \{\chi(g)S_g : g \in G\}$. Note that ${}_{\chi}\mathcal{S}$ is itself a bounded representation of G on \mathfrak{X} . Indeed, there is

$$\chi(g_1)S_{g_1}\chi(g_2)S_{g_2} = \chi(g_2)\chi(g_1)S_{g_1}S_{g_2} = \chi(g_2g_1)S_{g_2g_1}$$

and

$$\sup_{g \in G} \|\chi(g)S_g\| = \sup_{g \in G} |\chi(g)| \|S_g\| = \sup_{g \in G} \|S_g\| < \infty$$

Transformation $G \ni g \mapsto \chi(g)S_g x$ is continuous for a fixed $x \in \mathfrak{X}$ since

$$\begin{aligned} \|\chi(g)S_g x - \chi(h)S_h x\| &\leq \|\chi(g)S_g x - \chi(g)S_h x\| + \|\chi(g)S_h x - \chi(h)S_h x\| = \\ &= |\chi(g)| \|S_g x - S_h x\| + |\chi(g) - \chi(h)| \|S_h x\| \leq \|S_g x - S_h x\| + |\chi(g) - \chi(h)| \|x\| \sup_{q \in G} \|S_q\|, \end{aligned}$$

what can be made arbitrarily small when h is taken from a suitable open neighborhood of g , since χ and $g \mapsto S_g x$ are continuous on G .

Example 5.1 It is straightforward to check that a map $\chi : \mathbb{N} \rightarrow \mathbb{T}$ given by $\chi(n) := \lambda^n$ is a character of \mathbb{N} for a fixed $\lambda \in \mathbb{T}$. If now $\mathcal{S} = \{T_n\}_{n \in \mathbb{N}}$ is a bounded representation of \mathbb{N} on a Banach space \mathfrak{X} , then we have ${}_{\chi}\mathcal{S} = \{\lambda^n T_n : n \in \mathbb{N}\}$. The Cesaro averages $A_N^{{}_{\chi}\mathcal{S}}$ become

$$A_N^{{}_{\chi}\mathcal{S}} = \frac{1}{N} \sum_{n=0}^N \lambda^n T_n.$$

$A_N^{{}_{\chi}\mathcal{S}}$ forms a strong ${}_{\chi}\mathcal{S}$ -ergodic net (cf. Example 4.3). In particular, if T is a continuous transformation on the compact Hausdorff topological space X , then for the distorted semigroup ${}_{\chi}\mathcal{S} = \{\lambda^n U_T^n : n \in \mathbb{N}\}$, $\lambda \in \mathbb{T}$ induced by its Koopman operator U_T we have for any $f \in C(X)$

$$A_N^{{}_{\chi}\mathcal{S}} f = \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n U_T^n f = \frac{1}{N} \sum_{n=0}^{N-1} \lambda^n f \circ T^n,$$

hence Cesaro averages of distorted representation are exactly Wiener-Wintner type averages. We can of course take also Koopman operator on $L^p(\mu)$ ($1 \leq p < \infty$) of measure preserving transformation on measure space (X, \mathcal{A}, μ) .

Remark Let \mathcal{S} be a bounded Markovian representation of the (right) amenable semitopological semigroup G on $C(X)$. Given an invariant probability measure $\mu \in \mathbb{P}_{\mathcal{S}}$ we will consider the extended semigroup $\mathcal{S}_{2,\mu}$ and its distortion ${}_{\chi}\mathcal{S}_{2,\mu}$ for a given character $\chi \in \widehat{G}$. Note that \mathcal{S} , ${}_{\chi}\mathcal{S}$, $\mathcal{S}_{2,\mu}$, ${}_{\chi}\mathcal{S}_{2,\mu}$ are all bounded representations of (right) amenable semigroup, hence they all admit (right) strong ergodic net (cf. Theorem 4.4).

Even if \mathcal{S} is uniquely ergodic, the distorted semigroup ${}_{\chi}\mathcal{S}$ need not to be mean ergodic. We have seen that already in the Example 3.1. We will restate that example in the language of operator semigroups.

Example 5.2 (R. Derndinger, [Schreiber14, ex. 2.4]) Let (K, T) be a uniquely ergodic topological dynamical system such as in the Example 3.1. The semigroup $\mathcal{S} = \{U_T^n : n \in \mathbb{N}\}$ of Koopman operators on $C(K)$ is a bounded Markovian representation of \mathbb{N} on $C(K)$. We showed that $\mathbb{P}_{\mathcal{S}} = \{\frac{1}{2}(\delta_{\underline{x}(1)} + \delta_{-\underline{x}(1)})\}$, hence \mathcal{S} is uniquely and mean ergodic (Fact 4.3). We have seen that there exists $f \in C(K)$ for which the sequence $\frac{1}{N} \sum_{n=0}^{N-1} (-1)^n U_T^n f$ does not converge in $C(K)$.

For $\chi \in \widehat{G}$ given by $\chi(n) = (-1)^n$, the sequence $A_N^{{}_{\chi}\mathcal{S}} = \frac{1}{N} \sum_{n=0}^N \lambda^n U_T^n$ is a strong ${}_{\chi}\mathcal{S}$ -ergodic net on $C(K)$, hence the lack of the convergence shows that ${}_{\chi}\mathcal{S}$ is not mean ergodic. ■

5.2. Mean ergodicity of the semigroup ${}_{\chi}\mathcal{S}$

Let us state now Wiener-Wintner type result for uniquely ergodic semigroups of Markov operators. It formulates necessary and sufficient condition for mean ergodicity of ${}_{\chi}\mathcal{S}$ in terms of the semigroup ${}_{\chi}\mathcal{S}_{2,\mu}$, where μ is the unique invariant probability measure. The proof, different than in [Schreiber14], will be given later (as a corollary from Theorem 5.3).

Theorem 5.1 (M. Schreiber [Schreiber14, thm 2.7])

Let X be a compact Hausdorff topological space. Let \mathcal{S} be a bounded Markovian representation of a (right) amenable semitopological semigroup G on $C(X)$ and assume that \mathcal{S} is uniquely ergodic with invariant probability measure μ . The following conditions are equivalent:

- (1) $Fix(\chi\mathcal{S}_{2,\mu}) \subset Fix(\chi\mathcal{S})$
- (2) $A_\alpha^{\chi\mathcal{S}}x$ converges (to a fixed point of $\chi\mathcal{S}$) for every $x \in \mathfrak{X}$ and some/every strong (right) $\chi\mathcal{S}$ -ergodic net $(A_\alpha^{\chi\mathcal{S}})_{\alpha \in \Lambda}$,
- (3) $A_\alpha^{\chi\mathcal{S}}x$ converges weakly (to a fixed point of $\chi\mathcal{S}$) for every $x \in \mathfrak{X}$ and some/every strong (right) $\chi\mathcal{S}$ -ergodic net $(A_\alpha^{\chi\mathcal{S}})_{\alpha \in \Lambda}$,
- (4) $A_\alpha^{\chi\mathcal{S}}x$ converges weakly (to a fixed point of $\chi\mathcal{S}$) for every $x \in \mathfrak{X}$ and some/every weak (right) $\chi\mathcal{S}$ -ergodic net $(A_\alpha^{\chi\mathcal{S}})_{\alpha \in \Lambda}$,
- (5) $A_\alpha^{\chi\mathcal{S}}x$ has a weak cluster point (in $Fix(\chi\mathcal{S})$) for every $x \in \mathfrak{X}$ and some/every weak (right) $\chi\mathcal{S}$ -ergodic net $(A_\alpha^{\chi\mathcal{S}})_{\alpha \in \Lambda}$,
- (6) $\overline{co}_\chi \mathcal{S}x \cap Fix(\chi\mathcal{S}) \neq \emptyset$ for all $x \in \mathfrak{X}$,
- (7) there exists operator $P_\chi \in \overline{co}_\chi \mathcal{S}$ such that $\chi(g)S_gP = P\chi(g)S_g = P^2 = P$ for every $g \in G$ (i.e. $\chi\mathcal{S}$ is mean ergodic with the absorbing projection P_χ),
- (8) $Fix(\chi\mathcal{S})$ separates $Fix(\chi\mathcal{S})'$,
- (9) $\mathfrak{X} = Fix(\chi\mathcal{S}) \oplus \overline{span} \, rg(I - \chi\mathcal{S})$.

The strong/weak limit $\lim_{\alpha} A_\alpha^{\chi\mathcal{S}}x$ is equal for every strong/weak (right) $\chi\mathcal{S}$ -ergodic net $(A_\alpha^{\chi\mathcal{S}})_{\alpha \in \Lambda}$ and is also equal to Px . Operator P is a $\chi\mathcal{S}$ -absorbing projection onto $Fix(\chi\mathcal{S})$ along $\overline{span} \, rg(I - \chi\mathcal{S})$.

Remark The above theorem can be stated briefly: if \mathcal{S} is a bounded Markovian representation of a (right) amenable semitopological semigroup G on $C(X)$ then for every $\chi \in \widehat{G}$ the distorted semigroup $\chi\mathcal{S}$ is mean ergodic if and only if $Fix(\chi\mathcal{S}_{2,\mu}) \subset Fix(\chi\mathcal{S})$. Conditions (2) to (8) are equivalent formulations of the ergodicity of the semigroup $\chi\mathcal{S}$ and are consequence of the Abstract ergodic theorem (Theorems 4.2 and 4.3). The inclusion $Fix(\chi\mathcal{S}_{2,\mu}) \subset Fix(\chi\mathcal{S})$ is understood in $L^2(\mu)$, i.e. $Fix(\chi\mathcal{S}) \subset C(X)$ can be seen as a linear subspace of $L^2(\mu)$ (after identification of functions being equal almost everywhere). The inclusion says that for every function $f \in L^2(\mu)$ satisfying $f = \chi(g)S_g f$ μ -a.e. for all $g \in G$, there exists function $h \in C(X)$ with $f = h$ μ -a.e. satisfying $h = \chi(g)S_g h$ for all $g \in G$.

For a given $\mu \in \mathbb{P}_\mathcal{S}$ and $h \in L^2(\mu)$ let us denote by $h\mu$ a measure given by

$$h\mu(A) := \int_A h d\mu, \quad A \in \mathcal{B}(X).$$

We will use later the following

Fact 5.1 ([Rudin2, thm 6.12]) *For every complex-valued measure $\mu \in \mathcal{M}(X)$ there exists a measurable function $h : X \rightarrow \mathbb{C}$ with $|h(x)| = 1$ for every $x \in X$ such that $\mu = h|\mu|$.*

Lemma 5.1 ([Schreiber14, lem. 2.5])

Let X be a compact Hausdorff topological space. Let \mathcal{S} be a bounded Markovian representation of a right amenable semigroup G on $C(X)$ and take $\mu \in \mathbb{P}_\mathcal{S}$. For every $h \in Fix(\chi\mathcal{S}_{2,\mu})^$ the measure $\bar{h}\mu$ is $\chi\mathcal{S}$ -invariant, i.e. $\bar{h}\mu \in Fix(\chi\mathcal{S})'$. Conversely, if for some $h \in L^2(\mu)$ we have $\bar{h}\mu \in Fix(\chi\mathcal{S})'$, then $h \in Fix(\chi\mathcal{S}_{2,\mu})^*$.*

Proof: Let us take $h \in \text{Fix}(\chi\mathcal{S}_{2,\mu})^*$. For $f \in C(X)$, $g \in G$ we have

$$\begin{aligned} \langle f, \chi(g)S'_g(\bar{h}\mu) \rangle &= \langle \chi(g)S_g f, \bar{h}\mu \rangle = \int_X \chi(g)S_g f \bar{h} d\mu = \langle \chi(g)S_g f, h \rangle_{L^2(\mu)} = \langle \chi(g)S_{g,2} f, h \rangle_{L^2(\mu)} = \\ &= \langle f, \overline{\chi(g)}S_{g,2}^* h \rangle_{L^2(\mu)} = \langle f, (\chi(g)S_{g,2})^* h \rangle_{L^2(\mu)} = \langle f, h \rangle_{L^2(\mu)} = \int_X f \bar{h} d\mu = \langle f, \bar{h}\mu \rangle, \end{aligned}$$

so $\bar{h}\mu \in \text{Fix}(\chi\mathcal{S})'$. If we assume now that $\bar{h}\mu \in \text{Fix}(\chi\mathcal{S})'$ for some $h \in L^2(\mu)$, then for $f \in C(X)$ there is

$$\begin{aligned} \langle f, (\chi(g)S_{g,2})^* h \rangle_{L^2(\mu)} &= \langle \chi(g)S_{g,2} f, h \rangle_{L^2(\mu)} = \int_X \chi(g)S_{g,2} f \bar{h} d\mu = \int_X \chi(g)S_g f \bar{h} d\mu = \\ &= \langle \chi(g)S_g f, \bar{h}\mu \rangle = \langle f, \chi(g)S'_g \bar{h}\mu \rangle = \langle f, \bar{h}\mu \rangle = \langle f, h \rangle_{L^2(\mu)}. \end{aligned}$$

Since $C(X)$ is dense in $L^2(\mu)$, there is $\langle f, (\chi(g)S_{g,2})^* h \rangle_{L^2(\mu)} = \langle f, h \rangle_{L^2(\mu)}$ for every $f \in L^2(\mu)$, hence $h = (\chi(g)S_{g,2})^* h$ for every $g \in G$. \square

Fact 5.2 ([Schreiber14]) *Let H be a Hilbert space. If $T \in L(H)$ is a contraction then $\text{Fix}(T) = \text{Fix}(T^*)$.*

Proof: Note that $\|T^*\| = \|T\| \leq 1$. For every $x \in \text{Fix}(T)$ there is

$$\begin{aligned} \|T^*x - x\|^2 &= \langle T^*x - x, T^*x - x \rangle = \langle T^*x, T^*x \rangle - \langle T^*x, x \rangle - \langle x, T^*x \rangle + \langle x, x \rangle = \\ &= \|T^*x\|^2 - \langle x, Tx \rangle - \langle Tx, x \rangle + \|x\|^2 \leq \|x\|^2 - 2\|x\|^2 + \|x\|^2 = 0, \end{aligned}$$

so $x \in \text{Fix}(T^*)$, hence $\text{Fix}(T) \subset \text{Fix}(T^*)$. Since $(T^*)^* = T$, there also is $\text{Fix}(T^*) \subset \text{Fix}(T)$. \square

Remark The above fact implies that in Lemma 5.1 we can take h from $\text{Fix}(\chi\mathcal{S}_{2,\mu})$ instead of $\text{Fix}(\chi\mathcal{S}_{2,\mu})^*$.

We will give now two main theorems of this chapter.

Theorem 5.2

Let X be a compact Hausdorff topological space. Let \mathcal{S} be a bounded Markovian representation of a right amenable semigroup G on the Banach space $C(X)$. If \mathcal{S} is mean ergodic then for any character $\chi \in \widehat{G}$ the distorted semigroup $\chi\mathcal{S}$ is mean ergodic if and only if for every $\mu \in \mathbb{P}_{\mathcal{S}}$ there is $\overline{\text{Fix}(\chi\mathcal{S})}^{L^2(\mu)} = \text{Fix}(\chi\mathcal{S}_{2,\mu})$.

Proof: It follows from the Abstract mean ergodic theorem (Theorem 4.2) that it suffices to show equivalence between (1) $\overline{\text{Fix}(\chi\mathcal{S})}^{L^2(\mu)} = \text{Fix}(\chi\mathcal{S}_{2,\mu})$ and (2) $\text{Fix}(\chi\mathcal{S})$ separates $\text{Fix}(\chi\mathcal{S})'$.

(1) \Rightarrow (2) Let $\nu \in \text{Fix}(\chi\mathcal{S})'$ be nonzero and assume (without loss of generality) that $|\nu|$ is a probability measure. By Lemma 4.1.4 there is for every $g \in G$

$$S'_g|\nu| \geq |S'_g\nu| = |\overline{\chi(g)}\chi(g)S'_g\nu| = |\overline{\chi(g)}\nu| = |\nu|.$$

Lemma 4.1.6 yields now $S'_g|\nu| = |\nu|$, hence $|\nu| \in \mathbb{P}_{\mathcal{S}}$. By Fact 5.1 there exists measurable function $h : X \rightarrow \mathbb{C}$ with modulus 1 such that $\nu = \bar{h}|\nu|$. Since h has constant modulus, there

is $h \in L^2(|\nu|)$, hence by Lemma 5.1 and Fact 5.2 we have $h \in \text{Fix}(\chi \mathcal{S}_{2,|\nu|})$. By (1) we find a sequence $h_n \in \text{Fix}(\chi \mathcal{S})$ such that $h_n \xrightarrow{L^2(|\nu|)} h$. There is now

$$1 = \|h\|_{L^2(|\nu|)}^2 = \lim_{n \rightarrow \infty} \langle h_n, h \rangle_{L^2(|\nu|)} = \lim_{n \rightarrow \infty} \int_X h_n \bar{h} d|\nu| = \lim_{n \rightarrow \infty} \int_X h_n d\nu = \lim_{n \rightarrow \infty} \langle h_n, \nu \rangle,$$

so there is $\langle h_n, \nu \rangle \neq 0$ for some $n \in \mathbb{N}$. It follows that $\text{Fix}(\chi \mathcal{S})$ separates $\text{Fix}(\chi \mathcal{S})'$.

(2) \Rightarrow (1) Suppose that there exists $\mu \in \mathbb{P}_{\mathcal{S}}$ such that $\overline{\text{Fix}(\chi \mathcal{S})}^{L^2(\mu)} \neq \text{Fix}(\chi \mathcal{S}_{2,\mu})$. There always is $\text{Fix}(\chi \mathcal{S}) \subset \text{Fix}(\chi \mathcal{S}_{2,\mu})$ (since $\chi(g)S_g f = f$ implies $\chi(g)S_{g,2}f = f$ μ -a.e.) and $\text{Fix}(\chi \mathcal{S}_{2,\mu})$ is closed in $L^2(\mu)$, so there must be $\overline{\text{Fix}(\chi \mathcal{S})}^{L^2(\mu)} \subsetneq \text{Fix}(\chi \mathcal{S}_{2,\mu})$. Hence there exists $0 \neq f \in \text{Fix}(\chi \mathcal{S}_{2,\mu})$ with $f \perp \overline{\text{Fix}(\chi \mathcal{S})}^{L^2(\mu)}$. Lemma 5.1 and Fact 5.2 yield again that $\bar{f}\mu \in \text{Fix}(\chi \mathcal{S})'$. Since (2) there exists $h \in \text{Fix}(\chi \mathcal{S})$ such that $\langle g, \bar{f}\mu \rangle \neq 0$, but $\langle g, f \rangle_{L^2(\mu)} = \int_X g \bar{f} d\mu = \langle g, \bar{f}\mu \rangle \neq 0$, what contradicts $f \perp \text{Fix}(\chi \mathcal{S})$. \square

M. Schreiber showed in [Schreiber14] that if \mathcal{S} is uniquely ergodic with $\mathbb{P}_{\mathcal{S}} = \{\mu\}$, then $\dim(\text{Fix}(\chi \mathcal{S}_{2,\mu})) \leq 1$, so $\text{Fix}(\chi \mathcal{S}) \subset \text{Fix}(\chi \mathcal{S}_{2,\mu}) \subset L^2(\mu)$ is closed as a finite dimensional subspace, so the $L^2(\mu)$ -closure is redundant in the Theorem 5.2. We extend this fact to the case of mean ergodic Markov representations \mathcal{S} with finite dimensional $\text{Fix}(\mathcal{S})$ and give more elementary proof.

Theorem 5.3

Let X be a compact Hausdorff topological space. Let \mathcal{S} be a bounded Markovian representation of a right amenable semigroup G on the Banach space $C(X)$. If \mathcal{S} is mean ergodic with $\dim(\text{Fix}(\mathcal{S})) < \infty$, then then for any character $\chi \in \hat{G}$ the distorted semigroup $\chi \mathcal{S}$ is mean ergodic if and only if for every $\mu \in \mathbb{P}_{\mathcal{S}}$ there is $\text{Fix}(\chi \mathcal{S}) = \text{Fix}(\chi \mathcal{S}_{2,\mu})$.

For the proof of this theorem we will need the following geometrical lemma.

Lemma 5.2

Let (X, \mathcal{A}, μ) be a probability space. Let $f : X \rightarrow \mathbb{C}$ be such that $\mu(\{|f(x)| = c\}) = 1$ for some constant $c \geq 0$. If there also is $|\int_X f d\mu| = c$ then f is constant μ -a.e.

Proof: Assume that $c > 0$ (otherwise $f = 0$ μ -a.e.) and that f is not constant μ -a.e. Let us assume first that there exists $r \in \mathbb{R}$ with $\mu(\{\text{Re} f = r\}) = 1$. Note that there is $\{z \in \mathbb{R} : \text{Re} z = r, |z| = c\} = \{z_1, z_2\}$ for some (not necessarily distinct) complex numbers z_1, z_2 . There is $\mu(\{f = z_1 \text{ or } f = z_2\}) = \mu(\{|f(x)| = c\} \cap \{\text{Re} f = r\}) = 1$. Since f is not constant μ -a.e. there must be $0 < \mu(\{f = z_1\}) < 1$ and $0 < \mu(\{f = z_2\}) < 1$, but then

$$\int_X f d\mu = z_1 \cdot \mu(\{f = z_1\}) + z_2 \cdot \mu(\{f = z_2\}).$$

z_1 and z_2 were both from the disk $D_c := \{z \in \mathbb{C} : |z| \leq c\}$, so $\int_X f d\mu$ is not an extremat point of the set D_c . But there is $\text{Ext}(D_c) = \{z \in \mathbb{C} : |z| = c\}$, so there cannot be $|\int_X f d\mu| = c$. If there is no $r \in \mathbb{R}$ with $\mu(\{\text{Re} f = r\}) = 1$ then there mu exists $r \in \mathbb{R}$ with $0 < \mu(\{\text{Re} f \leq r\}) < 1$ and $0 < \mu(\{\text{Re} f > r\}) < 1$. There is now

$$\int_X f d\mu = \int_{\{\text{Re} f \leq r\}} f d\mu + \int_{\{\text{Re} f > r\}} f d\mu,$$

$$\left| \int_{\{\text{Ref} \leq r\}} f d\mu \right| \leq \int_{\{\text{Ref} \leq r\}} |f| d\mu = c \cdot \mu(\{\text{Ref} \leq r\})$$

and

$$\left| \int_{\{\text{Ref} > r\}} f d\mu \right| \leq \int_{\{\text{Ref} > r\}} |f| d\mu = c \cdot \mu(\{\text{Ref} > r\}),$$

hence

$$\frac{1}{\mu(\{\text{Ref} \leq r\})} \int_{\{\text{Ref} \leq r\}} f d\mu \in D_c \quad \text{and} \quad \frac{1}{\mu(\{\text{Ref} > r\})} \int_{\{\text{Ref} > r\}} f d\mu \in D_c.$$

There is also

$$\text{Re} \left(\int_{\{\text{Ref} \leq r\}} f d\mu \right) = \left(\int_{\{\text{Ref} \leq r\}} \text{Ref} d\mu \right) \leq r \cdot \mu(\{\text{Ref} \leq r\})$$

and

$$\text{Re} \left(\int_{\{\text{Ref} > r\}} f d\mu \right) = \left(\int_{\{\text{Ref} > r\}} \text{Ref} d\mu \right) > r \cdot \mu(\{\text{Ref} > r\}),$$

hence

$$\frac{1}{\mu(\{\text{Ref} \leq r\})} \int_{\{\text{Ref} \leq r\}} f d\mu \neq \frac{1}{\mu(\{\text{Ref} > r\})} \int_{\{\text{Ref} > r\}} f d\mu.$$

We received $\int_X f d\mu$ as a proper convex combination of complex numbers from D_c , hence $\int_X f d\mu$ cannot lie on a circle with radius c . \square

Proof: (of the Theorem 5.3) If for every $\mu \in \mathbb{P}_{\mathcal{S}}$ there is $\text{Fix}(\chi\mathcal{S}) = \text{Fix}(\chi\mathcal{S}_{2,\mu})$, then there also is $\overline{\text{Fix}(\chi\mathcal{S})}^{L^2(\mu)} = \text{Fix}(\chi\mathcal{S}_{2,\mu})$, so Theorem 5.2 gives that $\chi\mathcal{S}$ is mean ergodic. Assume conversly that $\chi\mathcal{S}$ is mean ergodic. We will show that $\text{Fix}(\chi\mathcal{S})$ is a finite dimensional subspace of $L^2(\mu)$ for every $\mu \in \mathbb{P}_{\mathcal{S}}$. Mean ergodicity of \mathcal{S} gives that $\text{Fix}(\mathcal{S})$ separates $\text{Fix}(\mathcal{S})'$ (Theorem 4.2) and $\text{Fix}(\mathcal{S}')$ separates $\text{Fix}(\mathcal{S})$ (Fact 4.1), hence $\dim(\text{Fix}(\mathcal{S}')) = \dim(\text{Fix}(\mathcal{S})) < \infty$ (Fact 4.2). Hence there is only finitely many extremal invariant probabilities $\mu_1, \mu_2, \dots, \mu_k \in \text{Ext}\mathbb{P}_{\mathcal{S}} \subset \text{Fix}(\mathcal{S}')$, since distinct extremal invariant probabilities are linearly independent. Indeed, support of the extremal invariant probability is a minimal set (Lemma 4.2.10), hence any pair of distinct extremal invariant probabilities must have disjoint supports (otherwise their nonempty intersection would be a proper invariant subset of these supports (Lemma 4.2.1)). For $f \in \text{Fix}(\chi\mathcal{S})$ we have $\chi(g)S_g f = f$, hence $S_g f = \overline{\chi(g)}f$ and further $S_g|f| \geq |S_g f| = |\overline{\chi(g)}f| = |f|$. Integrating with respect to μ_j , $j = 1, 2, \dots, k$ gives

$$\int_X S_g|f| d\mu_j = \langle S_g|f|, \mu_j \rangle = \langle |f|, S'_g \mu_j \rangle = \langle |f|, \mu_j \rangle = \int_X |f| d\mu_j,$$

so there must be $S_g|f| = |f|$, μ_j -a.e. Hence $S_g|f| = |f|$ on $\text{supp}(\mu_j)$ (Fact 2.4) and $|f|$ is constant on every $\text{supp}(\mu_j)$ (consider function $|f|_{\text{supp}(\mu_j)}$ invariant for $\mathcal{S}|_{\text{supp}(\mu_j)}$ and apply Lemma 4.2.8). Let us take arbitrary $x \in \text{supp}(\mu_j)$ and $g \in G$. We have

$$f(x) = \langle f, \delta_x \rangle = \langle \chi(g)S_g f, \delta_x \rangle = \chi(g) \langle f, S'_g \delta_x \rangle = \chi(g) \int_X f(y) dS'_g \delta_x(y) =$$

$$= \chi(g) \int_{\text{supp}(S'_g \delta_x)} f(y) dS'_g \delta_x(y).$$

Since f is of constant modulus on $\text{supp}(S'_g \delta_x) \subset \text{supp}(\mu_j)$, there is even $f(y) = \overline{\chi(g)} f(x)$ for every $y \in \text{supp}(S'_g \delta_x)$ (Lemma 5.2). If we have now $f_1, f_2 \in \text{Fix}_\chi(\mathcal{S})$ such that $f_2 \neq 0$ on $\text{supp}(\mu_j)$, then for $x \in \text{supp}(\mu_j)$, $g \in G$ there is

$$S_g\left(\frac{f_1}{f_2}\right)(x) = \int_X S_g\left(\frac{f_1}{f_2}\right)(y) d\delta_x(y) = \int_{\text{supp}(S'_g \delta_x)} \frac{f_1(y)}{f_2(y)} dS'_g \delta_x(y) = \frac{\overline{\chi(g)} f_1(x)}{\overline{\chi(g)} f_2(x)} = \frac{f_1}{f_2}(x),$$

hence $\frac{f_1}{f_2}$ is constant on $\text{supp}(\mu_j)$ and $f_1 = c \cdot f_2$ on $\text{supp}(\mu_j)$ for some $c \in \mathbb{C}$. For every $j = 1, 2, \dots, k$ let $f_j \in \text{Fix}_\chi(\mathcal{S})$ be such that $f_j(x) \neq 0$ for every $x \in \text{supp}(\mu_j)$ if such f_j exists. Otherwise every $f \in \text{Fix}_\chi(\mathcal{S})$ has $f|_{\text{supp}(\mu_j)} = 0$ (since f is of constant modulus on $\text{supp}(\mu_j)$) and we can take f_j to be zero on $\text{supp}(\mu_j)$. Let us take now $\mu \in \mathbb{P}_\mathcal{S}$. Since $\mathbb{P}_\mathcal{S} = \overline{co}^* \text{Ext} \mathbb{P}_\mathcal{S}$ by Krein-Milman theorem (Theorem 1.7), there is $\mu_\alpha \xrightarrow{w^*}$ for some net $(\mu_\alpha)_{\alpha \in \Lambda}$ with $\mu_\alpha \in co(\text{Ext} \mathbb{P}_\mathcal{S}) \subset \mathcal{M}^1(X)$ for every $\alpha \in \Lambda$. There is $\text{supp}(\mu_\alpha) \subset \bigcup_{j=1}^k \text{supp}(\mu_j)$, so Fact

2.4.7 yields that also $\text{supp}(\mu) \subset \bigcup_{j=1}^k \text{supp}(\mu_j)$. Hence every $f \in \text{Fix}_\chi(\mathcal{S})$ has decomposition

$f = \sum_{j=1}^k f \mathbf{1}_{\text{supp}(\mu_j)}$ in $L^2(\mu)$. Now $f \mathbf{1}_{\text{supp}(\mu_j)} = c_j f_j \mathbf{1}_{\text{supp}(\mu_j)}$ for some constant $c_j \in \mathbb{C}$, so $f = \sum_{j=1}^k c_j f_j \mathbf{1}_{\text{supp}(\mu_j)}$ in $L^2(\mu)$. This shows that $\dim(\text{Fix}_\chi(\mathcal{S})) \leq k$ in $L^2(\mu)$. \mathcal{S} is mean ergodic, so Theorem 5.2 gives that

$$\text{Fix}_\chi(\mathcal{S}_{2,\mu}) = \overline{\text{Fix}_\chi(\mathcal{S})}^{L^2(\mu)} = \text{Fix}_\chi(\mathcal{S}),$$

since finite dimensional subspace is always closed. \square

We can give now short proof of the Theorem 5.1.

Proof: (of the Theorem 5.1) If \mathcal{S} is uniquely ergodic with unique invariant probability measure $\mu \in \mathbb{P}_\mathcal{S}$, then by Fact 4.3 \mathcal{S} is also mean ergodic and $\text{Fix}(\mathcal{S}) = \mathbb{C} \cdot \mathbf{1}$, hence $\dim(\text{Fix}(\mathcal{S})) = 1 < \infty$. Theorem 5.3 states now that for every $\chi \in \widehat{G}$ the distorted semigroup $\chi \mathcal{S}$ is mean ergodic if and only if $\text{Fix}_\chi(\mathcal{S}_{2,\mu}) = \text{Fix}_\chi(\mathcal{S})$. The inclusion $\text{Fix}_\chi(\mathcal{S}_{2,\mu}) \supset \text{Fix}_\chi(\mathcal{S})$ is always true, hence the condition is equivalent to $\text{Fix}_\chi(\mathcal{S}_{2,\mu}) \subset \text{Fix}_\chi(\mathcal{S})$. \square

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