

On the asymptotic behaviour of some Markov processes

by

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The purpose of this paper is essentially to study the relations between the ergodicity of a Markov process and the ergodicity of Markov processes constructed from the former with the help of a measurable mapping (Part II). The conditions under which this construction is possible have been given in [5]. They are slightly modified here; this is the main object of the first part, to which we add some complements. The terminology used in this paper is that of [1] and [5].

In the following, we shall denote by N the set of non-negative integers, N^* the set of positive integers, and 1_B the indicator of the set B .

I. MARKOV PROCESSES CONSTRUCTED FROM ANOTHER WITH THE HELP OF A MEASURABLE MAPPING

I. 0. A σ -algebra \mathcal{T} of subsets of a non-empty set \mathcal{U} is said to be *separable* if there exists a countable family of subsets of \mathcal{U} generating \mathcal{T} . For example, if \mathcal{U} is a Polish space, and if \mathcal{T} is the σ -algebra generated by open sets of \mathcal{U} , then \mathcal{T} is separable.

A class \mathcal{C} of subsets of \mathcal{U} is said to be *semi-compact* if for every sequence $(C_n)_{n \in N^*}$ of elements of \mathcal{C} , such that $\bigcap_{n \in N^*} C_n = \emptyset$, there exists an $m \in N^*$ such that $\bigcap_{n \leq m} C_n = \emptyset$.

A class \mathcal{F} of subsets of \mathcal{U} is said to have the *property of approximation* relatively to a probability Q and to a σ -algebra \mathcal{T} if

$$\forall C \in \mathcal{T}, \quad Q(C) = \sup_{\substack{F \in \mathcal{F} \\ F \subset C}} Q(F).$$

I. 1. Let $(\mathcal{X}', \mathcal{B}')$ and $(\mathcal{X}'', \mathcal{B}'')$ be two measurable spaces. We recall that a transition probability P from $(\mathcal{X}', \mathcal{B}')$ to $(\mathcal{X}'', \mathcal{B}'')$ is a mapping $P: \mathcal{X}' \times \mathcal{B}'' \rightarrow [0, 1]$ such that $\forall x \in \mathcal{X}', P(x, \cdot)$ is a probability measure on \mathcal{B}'' , and $\forall B \in \mathcal{B}'', P(\cdot, B)$ is a real random variable defined on $(\mathcal{X}', \mathcal{B}')$. Let $((\mathcal{X}_t, \mathcal{B}_t))_{t \in N^*}$ be a sequence of measurable spaces, $(P_{t, t+1})_{t \in N^*}$ a sequence of transition probabilities such that $\forall t \in N^*, P_{t, t+1}$ is a transition

probability from $(\mathcal{X}_t, \mathcal{B}_t)$ to $(\mathcal{X}_{t+1}, \mathcal{B}_{t+1})$. We shall denote by $P_{s,t}$ the transition probability from $(\mathcal{X}_s, \mathcal{B}_s)$ to $(\mathcal{X}_t, \mathcal{B}_t)$ defined by

$$P_{s,t}(x, B) = \int_{\mathcal{X}_{s+1}} P_{s,s+1}(x, dx_{s+1}) \dots \int_{\mathcal{X}_{t-1}} P_{t-2,t-1}(x_{t-2}, dx_{t-1}) P_{t-1,t}(x, B),$$

$$\forall (x, B) \in \mathcal{X}_s \times \mathcal{B}_t.$$

The family $((\mathcal{X}_t, \mathcal{B}_t), P_{t,t+1})_{t \in \mathbb{N}^*}$ is said to be a *Markov process*. In the special case $\mathcal{X}_t = \mathcal{X}$, $\mathcal{B}_t = \mathcal{B}$ and $P_{t,t+1} = P$, $\forall t \in \mathbb{N}^*$, we say that the Markov process is *homogeneous*, and we denote it by $((\mathcal{X}, \mathcal{B}), P_n)_{n \in \mathbb{N}^*}$, where $P_n = P_{s,s+n}$, $\forall s \in \mathbb{N}^*$.

Let $(\Omega, \mathcal{A}, \text{Pr})$ be a probability space. A Markov random function $(X_t)_{t \in \mathbb{N}^*}$ defined on $(\Omega, \mathcal{A}, \text{Pr})$ and taking its values in a measurable space $(\mathcal{X}, \mathcal{B})$ is said to be *attached* to the Markov process $((\mathcal{X}, \mathcal{B}), P_n)_{n \in \mathbb{N}^*}$ if it has P as transition probability.

Let $(\mathcal{Y}, \mathcal{T})$ be a measurable space and f be a \mathcal{T} - \mathcal{B} -measurable mapping from \mathcal{X} onto \mathcal{Y} . Let us denote by \mathcal{B}_f the sub- σ -algebra $f^{-1}(\mathcal{T})$ of \mathcal{B} . As mentioned in the introduction, the purpose of this part is to recall and to modify, slightly, results given by Rosenblatt in [5]. We also give some complements, which will be useful for the second part.

For convenience, we shall define the following hypotheses:

I. 2. (i) $\forall x \in \mathcal{X}$, $\{x\} \in \mathcal{B}$,

(ii) $\forall y \in \mathcal{Y}$, $\{y\} \in \mathcal{T}$,

(iii) $\forall x \in \mathcal{X}$, $P_{\mathcal{B}_f}(x, \cdot)$ is dominated (viz. absolutely continue with respect to) by a positive σ -finite measure μ on \mathcal{B}_f , $P_{\mathcal{B}_f}(x, \cdot)$ being the restriction of $P(x, \cdot)$ to \mathcal{B}_f .

I. 3. \mathcal{T} is separable and there exists a semi-compact sub-class \mathcal{F} of \mathcal{T} such that, for any probability Q on \mathcal{T} , \mathcal{F} has the property of approximation relatively to Q and to \mathcal{T} .

Proofs of propositions I. 10 and I. 11 below will be omitted: they are the same as those given in [5], modulo the following result:

I. 4. PROPOSITION. Let $A \in \mathcal{B}_f$, let ξ be a probability on \mathcal{B} and ξ_A be the restriction of ξ on \mathcal{B}_f . Under hypothesis I. 3, there exists a mapping $v: A \times \mathcal{T} \rightarrow [0, 1]$ such that

(i) $\forall C \in \mathcal{T}$, $v(\cdot, C)$ is $\mathcal{B}_f^{(A)}$ -measurable and $\forall B \in \mathcal{B}_f^{(A)}$,

$$\int_B v(x, C) d\xi(x) = \int_B P(x, f^{-1}(C)) d\xi(x),$$

$\mathcal{B}_f^{(A)}$ denoting the σ -algebra $\mathcal{B}_f \cap A$.

(ii) $\forall x \in A$, $v(x, \cdot)$ is a probability on \mathcal{T} .

We omit the proof. In the particular case $A = \mathcal{X}$, we have the

I. 5. COROLLARY. Under hypothesis I. 3, there exists a mapping $v: \mathcal{X} \times \mathcal{T} \rightarrow [0, 1]$ such that

(i) $\forall C \in \mathcal{T}$, $v(\cdot, C)$ is a version of the conditional expectation $E_\xi[P(\cdot, f^{-1}(C)) | \mathcal{B}_f]$ relatively to ξ .

(ii) $\forall x \in \mathcal{X}$, $v(x, \cdot)$ is a probability on \mathcal{T} .

I. 6. We know (cf. [3]) that: For every family of probabilities dominated by a probability, one can find an equivalent countable sub-family of probabilities. Since the family $(P_{\mathcal{B}_f}(x, \cdot))_{x \in \mathcal{X}}$ of probabilities is dominated, it is equivalent to a countable sub-family $(P_{\mathcal{B}_f}(x_i, \cdot))_{i \in \mathbb{N}^*}$. The dominating (in fact, equivalent) measure can be taken, and shall be taken, as the probability measure

$$\mu = \sum_{i \in \mathbb{N}^*} \alpha_i P_{\mathcal{B}_f}(x_i, \cdot), \quad \text{where } \alpha_i > 0 \text{ and } \sum_{i \in \mathbb{N}^*} \alpha_i = 1.$$

We shall also make use, in Part II, of the probability measure

$$\bar{\mu} = \sum_{i \in \mathbb{N}^*} \alpha_i P(x_i, \cdot) \quad \text{on } \mathcal{B}.$$

I. 7. A set $S \in \mathcal{B}_f$ is said to be a *single entry set* if $\mu(S) > 0$ and if there exists a point $y \in \mathcal{Y}$ such that $P(x, S) = 0$, $\forall x \in \mathbb{C} f^{-1}(y)$. Such a point y is called a *single entry point* corresponding to S . A set $S \in \mathcal{B}_f$ is said to be a *maximal single entry set* if it is a single entry set and if there does not exist any single entry set S' such that $S' \supset S$ and $\mu(S') > \mu(S)$.

I. 8. Let us now recall some results:

1° If S is a single entry set, there exists at least one point $x \in \mathcal{X}$ such that $P(x, S) > 0$.

2° A maximal single entry set S is defined modulo a μ -null set.

3° There are at most a countable class of disjoint maximal single entry sets $(S_k)_{k \in \mathbb{N}^*}$ and corresponding entry points.

4° Define $M = \mathbb{C} \left(\bigcup_{k \in \mathbb{N}^*} S_k \right)$. If $\mu(M) = 0$, then \mathcal{X} is the union of an at most countable family of maximal single entry sets.

5° If $\mu(M) = 0$ and if S is a maximal single entry set such that $P(x, S) > 0$, then $P(x, S) = 1$.

6° If $\mu(M) = 0$, and if $y \in \mathcal{Y}$ is such that $\mu(f^{-1}(y)) > 0$, then $f^{-1}(y)$ is a single entry set.

7° If $\mu(M) = 0$, then \mathcal{Y} is the at most countable set of single entry points.

I. 9. It is to be remarked that if $(X_t)_{t \in \mathbb{N}^*}$ is a Markov random function attached to a homogeneous Markov process $((\mathcal{X}, \mathcal{B}), P_n)_{n \in \mathbb{N}^*}$, the random function $(f \circ X_t)_{t \in \mathbb{N}^*}$ is not generally Markovian. Rosenblatt in [5] has given the conditions under which $(f \circ X_t)_{t \in \mathbb{N}^*}$ is also Markovian.

I. 10. PROPOSITION. Under hypothesis I. 2, and if $\mu(M) = 0$, then whatever be the Markov random function $(X_t)_{t \in \mathbb{N}^*}$ attached to the Markov process $((\mathcal{X}, \mathcal{B}), P_n)_{n \in \mathbb{N}^*}$, the random function $(f \circ X_t)_{t \in \mathbb{N}^*}$ is Markovian.

I. 11. Consider now the case $0 < \mu(M) < 1$. There exist some single entry sets, for this existence is equivalent to the condition $\mu(M) < 1$. Let S be a single entry set such that we may be able to enter S starting from M , that is to say there exist a positive integer n and n points $y_1, \dots, y_n \in \mathcal{Y}$ such that: a) $f^{-1}(y_j)$, $j = 2, \dots, n$, are single entry sets and $f^{-1}(y_1)$ is not; b) y_{j-1} is the single entry point corresponding to $f^{-1}(y_j)$, $j = 2, \dots, n$ and y_n is the single entry point corresponding to S . The integer n and the points y_1, \dots, y_n , when they exist, are uniquely defined and constitute a thread of length n entering S .

I. 12. PROPOSITION. Under the hypotheses I. 2 and I. 3, a necessary and sufficient condition for $(f \circ X_t)_{t \in \mathbb{N}^*}$ to be Markovian whatever be the Markov random function $(X_t)_{t \in \mathbb{N}^*}$, attached to the Markov process $((\mathcal{X}, \mathcal{B}), P_n)_{n \in \mathbb{N}^*}$, is that:

(i) There exists a mapping $R: M \times \mathcal{T} \rightarrow [0, 1]$ such that $\forall C \in \mathcal{T}$, $\forall x_0 \in \mathcal{X}$, $R(\cdot, C)$ is $\mathcal{B}_f^{(M)}$ -measurable and $\forall B \in \mathcal{B}_f^{(M)}$,

$$\int_B P(x_0, dx_1) R(x_1, C) = \int_B P(x_0, dx_1) P(x_1, f^{-1}(C)).$$

(ii) If S is a single entry set and if there exists a thread of length n entering S , then there exists a mapping $R_n: S \times \mathcal{T} \rightarrow [0, 1]$ such that $\forall C \in \mathcal{T}$, $\forall x_0 \in \mathcal{X}$, $R_n(\cdot, C)$ is $\mathcal{B}_f^{(S)}$ -measurable and $\forall B \in \mathcal{B}_f^{(S)}$,

$$\int_B P_{n+1}(x_0, dx_1) R_n(x_1, C) = \int_B P_{n+1}(x_0, dx_1) P(x_1, f^{-1}(C)).$$

As mentioned above, we omit the proof of I.12 and refer to [5]. We simply recall that the absolute probabilities \tilde{Q}_t and the transition probabilities $Q_{t,t+1}(\cdot, \cdot)$, $t \in \mathbb{N}^*$, of the Markov random function $(f \circ X_t)_{t \in \mathbb{N}^*}$ are constructed as follows:

1° π_t being the law of X_t , $t \in \mathbb{N}^*$, \tilde{Q}_t is defined by $\tilde{Q}_t(C) = \pi_t[f^{-1}(C)]$, $\forall C \in \mathcal{T}$.

2° With regard to the transition probabilities:

If $t = 1$, we define, $\forall y \in \mathcal{Y}$ and $\forall C \in \mathcal{T}$, $Q_{1,2}(y, C) = \nu_1(\cdot, C)$, where $x \in f^{-1}(y)$ and where $\nu_t(\cdot, C)$ is a version of the conditional expectation $E_{\pi_t}[P(\cdot, f^{-1}(C)) | \mathcal{B}_f]$ which is such that $\forall x \in \mathcal{X}$, $\nu_t(x, \cdot)$ is a probability on \mathcal{T} (such a version exists following I. 5).

If $t \geq 2$, we distinguish two cases:

a) If $y \in f(M)$, then $f^{-1}(y) \subset M$ (for $f^{-1}(y)$ is an atom of \mathcal{B}_f) and we define

$$Q_{t,t+1}(y, C) = R(x, C), \quad \forall C \in \mathcal{T}, \quad \text{where } x \in f^{-1}(y).$$

b) If $y \notin f(M)$, then $f^{-1}(y)$ is included in a maximal single entry set S . If there exists a thread of length n entering S , we put, $\forall C \in \mathcal{T}$,

$$Q_{t,t+1} = \begin{cases} R_n(x, C) & \text{if } t > n, \text{ where } x \in f^{-1}(y), \\ \nu_t(x, C) & \text{if } t \leq n, \text{ where } x \in f^{-1}(y), \end{cases}$$

$\nu_t(\cdot, C)$ denoting a $\mathcal{B}_f^{(S)}$ -measurable function such that $\forall B \in \mathcal{B}_f^{(S)}$,

$$\int_B d\pi_t(x) \nu_t(x, C) = \int_B d\pi_t(x) P(x, f^{-1}(C))$$

and $\forall x \in \mathcal{X}$, $\nu(\cdot, C)$ is a probability on \mathcal{T} (such a function exists by I. 3).

I. 13. Let us now make a few remarks:

1° $\forall t \geq 2$, $\forall B \in \mathcal{B}_f^{(M)}$ and $\forall C \in \mathcal{T}$,

$$(I. 13.1) \quad \int_B d\pi_t(x_1) P(x_1, f^{-1}(C)) = \int_B d\pi_t(x_1) R(x_1, C).$$

Indeed,

$$\begin{aligned} \int_B d\pi_t(x_1) P(x_1, f^{-1}(C)) &= \int_B \left[\int_{\mathcal{X}} d\pi_{t-1}(x) P(x, dx_1) \right] P(x_1, f^{-1}(C)) \\ &= \int_{\mathcal{X}} d\pi_{t-1}(x) \left[\int_B P(x, dx_1) P(x_1, f^{-1}(C)) \right] \\ &= \int_{\mathcal{X}} d\pi_{t-1}(x) \left[\int_B P(x, dx_1) R(x_1, C) \right] \\ &= \int_B \left[\int_{\mathcal{X}} d\pi_{t-1}(x) P(x, dx_1) \right] R(x_1, C) \\ &= \int_B d\pi_t(x_1) R(x_1, C). \end{aligned}$$

If $f^{-1}(y) \in \mathcal{B}_f^{(M)}$ and if $\pi_t[f^{-1}(y)] > 0$, then

$$(I. 13.2) \quad R(x, C) = \frac{1}{\pi_t[f^{-1}(y)]} \int_{f^{-1}(y)} d\pi_t(x_1) P(x_1, f^{-1}(C)), \quad \forall x \in f^{-1}(y).$$

Indeed, $R(\cdot, C)$ is constant on $f^{-1}(y)$ which is an atom of $\mathcal{B}_f^{(M)}$. We then make use of (I.13.1) by taking $f^{-1}(y)$ in place of B .

2° Similarly, we have: $\forall t \geq n+2$, $\forall B \in \mathcal{B}_f^{(S)}$ and $\forall C \in \mathcal{T}$,

$$(I. 13.3) \quad \int_B d\pi_t(x_1) P(x_1, f^{-1}(C)) = \int_B d\pi_t(x_1) R_n(x_1, C).$$

If $f^{-1}(y) \in \mathcal{B}_f^{(S)}$ and if $\pi_t[f^{-1}(y)] > 0$, then

$$(I. 13.4) \quad R_n(x, C) = \frac{1}{\pi_t[f^{-1}(y)]} \int_{f^{-1}(y)} d\pi_t(x_1) P(x_1, f^{-1}(C)), \quad \forall x \in f^{-1}(y).$$

3° If $\forall x_0 \in \mathcal{X}$, $\forall B \in \mathcal{B}_f$ and $\forall C \in \mathcal{T}$,

$$(I. 13.5) \quad \int_B P(x_0, dx_1) P(x_1, f^{-1}(C)) = \int_B P(x_0, dx_1) R(x_1, C),$$

where $R(\cdot, C)$ is \mathcal{B}_f -measurable. Then, we have $\forall x_0 \in \mathcal{X}, \forall B \in \mathcal{B}_f, \forall C \in \mathcal{T}$ and $\forall n \in \mathbb{N}^*$,

$$\begin{aligned}
 (I.13.6) \quad & \int_B P_{n+1}(x_0, dx_1) P(x_1, f^{-1}(C)) \\
 &= \int_B \left[\int_{\mathcal{X}} P_n(x_0, dx) P(x, dx_1) \right] P(x_1, f^{-1}(C)) \\
 &= \int_{\mathcal{X}} P_n(x_0, dx) \left[\int_B P(x, dx_1) P(x_1, f^{-1}(C)) \right] \\
 &= \int_{\mathcal{X}} P_n(x_0, dx) \left[\int_B P(x, dx_1) R(x_1, C) \right] \\
 &= \int_B P_{n+1}(x_0, dx_1) R(x_1, C).
 \end{aligned}$$

So, (I.13.5) implies I.12 (ii): it is sufficient to take $R_n = R$.

4° A sufficient condition for I.12 (i) and I.12 (ii) to be satisfied is that $P(\cdot, f^{-1}(C))$ is \mathcal{B}_f -measurable, $\forall C \in \mathcal{T}$.

Indeed, (I.13.5) is then verified, with $P(\cdot, f^{-1}(C))$ in place of $R(\cdot, C)$.

I.14. Consider now the homogeneous case. $(X_t)_{t \in \mathbb{N}^*}$ and $(f \circ X_t)_{t \in \mathbb{N}^*}$ having the same meanings as above, it is known that $((\mathcal{Y}, \mathcal{T}), Q_{t,t+1})_{t \in \mathbb{N}^*}$ is not necessarily homogeneous, although $((\mathcal{X}, \mathcal{B}), P_n)_{n \in \mathbb{N}^*}$ is.

A sufficient condition for $((\mathcal{Y}, \mathcal{T}), Q_{t,t+1})_{t \in \mathbb{N}^*}$ to be homogeneous is that $P(\cdot, B)$ is \mathcal{B}_f -measurable, $\forall B \in \mathcal{B}_f$.

In the case where there exists no single entry set, this sufficient condition is also necessary.

I.15. The following remarks will be useful for Part II.

1° Suppose that I.12 (i) and (ii) are verified. Let $v_{s,t}$ be defined by $v_{s,t}(x, C) = Q_{s,t}(y, C)$ for $y \in \mathcal{B}$, $x \in f^{-1}(y)$ and $C \in \mathcal{T}$. Then $v_{s,t}(\cdot, C)$ is a version of the conditional expectation $E_{\pi_s}[P_{t-s}(\cdot, f^{-1}(C)) | \mathcal{B}_f]$.

Indeed, $\forall C$ and $C' \in \mathcal{T}$, and $\forall s, t \in \mathbb{N}^*$ with $s < t$, we have

$$\Pr\{Y_s \in C', Y_t \in C\} = \Pr\{X_s \in f^{-1}(C'), X_t \in f^{-1}(C)\},$$

that is to say

$$(I.15.1) \quad \int_{C'} d\tilde{Q}_s(y) Q_{s,t}(y, C) = \int_{f^{-1}(C')} d\pi_s(x) P_{t-s}(x, f^{-1}(C)).$$

Since $v_{s,t}(\cdot, C)$ is \mathcal{B}_f -measurable, we have

$$\int_{C'} d\tilde{Q}_s(y) Q_{s,t}(y, C) = \int_{f^{-1}(C')} d\pi_{s, \mathcal{B}_f}(x) v_{s,t}(x, C),$$

π_{s, \mathcal{B}_f} denoting the restriction of π_s to \mathcal{B}_f . Then

$$\int_{f^{-1}(C')} d\pi_{s, \mathcal{B}_f}(x) v_{s,t}(x, C) = \int_{f^{-1}(C')} d\pi_s(x) P_{t-s}(x, f^{-1}(C)).$$

2° Suppose that (I.13.5) is verified. Let $R'(x, f^{-1}(C)) = R(x, C)$, $C \in \mathcal{T}$; $(R'(x, \cdot))$ is then a positive measure on \mathcal{B}_f and

$$R^{(1)}(x, C) = R(x, C),$$

$$R^{(n)}(x, C) = \int_{\mathcal{X}} R'(x, dx') R^{(n-1)}(x', C).$$

Then $\forall B \in \mathcal{B}_f, \forall t \in \mathbb{N}^*$ and $\forall (x_0, C) \in \mathcal{X} \times \mathcal{T}$,

$$(I.15.2) \quad \int_B P(x_0, dx) P_t(x, f^{-1}(C)) = \int_B P(x_0, dx) R^{(t)}(x, C).$$

We proceed by induction. Suppose that the equality is true for $n-1$, then $\forall B \in \mathcal{B}_f$,

$$\begin{aligned}
 \int_B P(x_0, dx) P_n(x, f^{-1}(C)) &= \int_B P(x_0, dx) \left[\int_{\mathcal{X}} P(x, dx') P_{n-1}(x', f^{-1}(C)) \right] \\
 &= \int_B P(x_0, dx) \left[\int_{\mathcal{X}} P(x, dx') R^{(n-1)}(x', C) \right] \\
 &= \int_{\mathcal{X}} \left[\int_B P(x_0, dx) P(x, dx') \right] R^{(n-1)}(x', C) \\
 &= \int_{\mathcal{X}} \left[\int_B P(x_0, dx) R'(x, dx') \right] R^{(n-1)}(x', C) \\
 &= \int_B P(x_0, dx) R^{(n)}(x, C).
 \end{aligned}$$

3° $\forall s \geq 2, \forall t > s$ and $\forall C \in \mathcal{T}$, we have μ -a.s.

$$(I.15.3) \quad Q_{s,t}(y, C) = R^{(t-s)}(x, C), \quad x \in f^{-1}(y).$$

At first, from I.15.2, we have $\forall B \in \mathcal{B}_f$,

$$\int_B d\bar{\mu}(x) P_n(x, f^{-1}(C)) = \int_B d\bar{\mu}(x) R^{(n)}(x, C) = \int_B d\mu(x) R^{(n)}(x, C),$$

for $R^{(n)}(\cdot, C)$ is \mathcal{B}_f -measurable.

Analogously, as in (I.13.3), we have: $\forall B \in \mathcal{B}_f, \forall s \geq 2, \forall t > s$ and $\forall C \in \mathcal{T}$,

$$\begin{aligned}
 \int_B d\pi_s(x) P_{t-s}(x, f^{-1}(C)) &= \int_B d\pi_s(x) R^{(t-s)}(x, C) \\
 &= \int_B d\pi_{s, \mathcal{B}_f}(x) R^{(t-s)}(x, C).
 \end{aligned}$$

Hence, by I.15.1°,

$$v_{s,t}(\cdot, C) = R^{(t-s)}(\cdot, C), \quad \pi_{s, \mathcal{B}_f} \text{-a.s.}$$

and then, π_{s, \mathcal{B}_f} -a.s.

$$Q_{s,t}(y, C) = R^{(t-s)}(x, C), \quad x \in f^{-1}(y).$$

On the other hand, $\forall s \geq 2, \pi_{s, \mathcal{B}_f}$ is dominated by μ , so that we have, μ -a.s.

$$Q_{s,t}(y, C) = R^{(t-s)}(x, C), \quad x \in f^{-1}(y).$$

II. PROBLEMS OF ERGODICITY

II. 1. We recall the following definitions (cf. [1]): A Markov process $(A, \mathcal{L}, q_{t,t+1})_{t \in \mathbb{N}^*}$ is said to be:

(i) *strongly ergodic* if $\forall s \in \mathbb{N}^*, \forall z \in A$ and $\forall A \in \mathcal{L}, \lim_{t \rightarrow \infty} q_{s,t}(z, A) = \xi_s(A)$, where ξ_s is a probability on \mathcal{L} ;

(ii) *weakly ergodic* if $\forall s \in \mathbb{N}^*, \forall z$ and $z' \in A$ and $\forall A \in \mathcal{L}, \lim_{t \rightarrow \infty} [q_{s,t}(z, A) - q_{s,t}(z', A)] = 0$;

(iii) *strongly (resp. weakly) and uniformly ergodic* if it is strongly (resp. weakly) ergodic and if, moreover, the limit in (i) (resp. (ii)) holds uniformly with respect to z and A .

II. 2. HYPOTHESIS. We suppose that (I. 13.5) is verified, that is to say: there exists a mapping $R: \mathcal{X} \times \mathcal{T} \rightarrow [0, 1]$ such that $\forall C \in \mathcal{T}, R(\cdot, C)$ is \mathcal{B}_f -measurable and that

$$\int_B P(x_0, dx_1) P(x_1, f^{-1}(C)) = \int_B P(x_0, dx_1) R(x_1, C),$$

$\forall x_0 \in \mathcal{X}, \forall C \in \mathcal{T}$ and $\forall B \in \mathcal{B}_f$.

By I.13.3°, we see that I.12 (ii) is verified.

II. 3. $\forall C \in \mathcal{T}, R(\cdot, C)$ is a version of $E_{P(x_0, \cdot)}[P(\cdot, f^{-1}(C)) | \mathcal{B}_f]$, then $R(\cdot, C)$ is also a version of $E_{\bar{\mu}}[P(\cdot, f^{-1}(C)) | \mathcal{B}_f]$. We have seen (Part I) that we can choose R in such a way that $\forall x \in \mathcal{X}, R(x, \cdot)$ is a probability on \mathcal{T} , and that for $s \geq 2, \mu$ -a.s.,

$$Q_{s,t}(y, C) = R^{(t-s)}(x, C), \quad x \in f^{-1}(y),$$

where $R^{(t-s)}(\cdot, C)$ is a version of $E_{\bar{\mu}}[P_{t-s}(\cdot, f^{-1}(C)) | \mathcal{B}_f]$.

More precisely, there exists a set $N_f \in \mathcal{B}_f$ (viz. there exists $N \in \mathcal{T}$ such that $N_f = f^{-1}(N)$), with $\mu(N_f) = 0$, such that $\forall s \geq 2, \forall t \geq s, \forall C \in \mathcal{T}$,

$$Q_{s,t}(y, C) = R^{(t-s)}(x, C), \quad x \in f^{-1}(y), \quad \text{if } y \notin N.$$

We specially consider the case where, $\forall s \geq 2$,

$$(II.3.1) \quad \begin{aligned} Q_{s,t}(y, C) &= R^{(t-s)}(x, C), & x \in f^{-1}(y), & \text{if } y \notin N; \\ Q_{s,s+1}(y, C) &= \mu'(C) & \text{if } y \in N, \end{aligned}$$

μ' denoting the image of μ by f .

II. 4. PROPOSITION. Under Hypothesis II.2, if $(X_t)_{t \in \mathbb{N}^*}$ is a Markov random function attached to a Markov process $((\mathcal{X}, \mathcal{B}), P_n)_{n \in \mathbb{N}^*}$ strongly and uniformly ergodic, then $(f \circ X_t)_{t \in \mathbb{N}^*}$ is a Markov random function attached to a strongly and uniformly ergodic Markov process.

Proof. 1° By hypothesis, $\lim_{t \rightarrow \infty} P_t(x, A) = \xi(A)$ uniformly with respect to x and A , where ξ is a probability on \mathcal{B} . Let $\mathcal{T}_0 = (C_0^t)_{t \in \mathbb{N}^*}$ be an algebra generating \mathcal{T} . $\forall C_0^t \in \mathcal{T}_0, R^{(t)}(\cdot, C_0^t)$ is a version of the conditional expectation $E_{\bar{\mu}}[P_t(\cdot, f^{-1}(C_0^t)) | \mathcal{B}_f]$. Since $0 \leq P_t(\cdot, f^{-1}(C_0^t)) \leq 1, \forall t \in \mathbb{N}^*$, we have

$$\lim_{t \rightarrow \infty} R^{(t)}(\cdot, C_0^t) = E_{\bar{\mu}}[\lim_{t \rightarrow \infty} P_t(\cdot, f^{-1}(C_0^t)) | \mathcal{B}_f] = \xi[f^{-1}(C_0^t)], \quad \mu\text{-a.s.}$$

But \mathcal{T}_0 is countable, so that there exists a set $N_1 \in \mathcal{B}_f$ such that $\mu(N_1) = 0$ and

$$(II.4.1) \quad \lim_{t \rightarrow \infty} R^{(t)}(x, C_0^t) = \xi[f^{-1}(C_0^t)], \quad \forall x \notin N_1 \text{ and } \forall C_0^t \in \mathcal{T}_0.$$

Let us now prove that this limit is uniform, μ -a.s. Indeed, by hypothesis, $\forall \varepsilon > 0, \exists n_\varepsilon \in \mathbb{N}^*$ such that $t > n_\varepsilon$ implies

$$-\varepsilon + \xi[f^{-1}(C_0^t)] < P_t(x, f^{-1}(C_0^t)) < \varepsilon + \xi[f^{-1}(C_0^t)], \quad \forall (x, C_0^t) \in \mathcal{X} \times \mathcal{T}_0.$$

Hence, $\forall B \in \mathcal{B}_f$,

$$(-\varepsilon + \xi[f^{-1}(C_0^t)]) \cdot \mu(B) < \int_B d\mu(x) P_t(x, f^{-1}(C_0^t)) < (\varepsilon + \xi[f^{-1}(C_0^t)]) \cdot \mu(B),$$

and, consequently, $\forall B \in \mathcal{B}_f$,

$$(-\varepsilon + \xi[f^{-1}(C_0^t)]) \cdot \mu(B) < \int_B d\mu(x) R^{(t)}(x, C_0^t) < (\varepsilon + \xi[f^{-1}(C_0^t)]) \cdot \mu(B).$$

This implies that

$$(II.4.2) \quad -\varepsilon + \xi[f^{-1}(C_0^t)] < R^{(t)}(\cdot, C_0^t) < \varepsilon + \xi[f^{-1}(C_0^t)], \quad \mu\text{-a.s.}$$

By (II.4.1) and (II.4.2), there exists a μ -null set $N_2 \in \mathcal{B}_f$ such that

$$(II.4.3) \quad \lim_{t \rightarrow \infty} R^{(t)}(x, C_0^t) = \xi[f^{-1}(C_0^t)], \quad \forall x \notin N_1 \cup N_2 \text{ and } \forall C_0^t \in \mathcal{T}_0$$

and this limit is uniform with respect to (x, C_0^t) on $\mathcal{C}(N_1 \cup N_2) \times \mathcal{T}_0$.

2° Let us prove now that (II.4.3) is verified for every $C \in \mathcal{T}$. For this purpose, let us prove that the class \mathcal{T}' of sets $C \in \mathcal{T}$ such that (II.4.3) is verified, is a monotone class containing \mathcal{T}_0 (and, consequently, \mathcal{T}' contains \mathcal{T}).

Indeed, it is clear that \mathcal{T}' contains \mathcal{T}_0 . It remains to prove that \mathcal{T}' is a monotone class. Let $(C_i)_{i \in \mathbb{N}^*}$ be a monotone sequence of elements of \mathcal{T} whose limit is C . For every $x \notin N_1 \cup N_2$, we have

$$(II.4.4) \quad \lim_{t \rightarrow \infty} R^{(t)}(x, C) = \lim_{t \rightarrow \infty} \lim_{i \rightarrow \infty} R^{(t)}(x, C_i),$$

for $R^{(t)}(x, \cdot)$ is a probability on \mathcal{T} .

Since the limit $\lim_{t \rightarrow \infty} R^{(t)}(x, C_i)$ is uniform with respect to C_i , we have

$$(II.4.5) \quad \lim_{t \rightarrow \infty} R^{(t)}(x, C) = \lim_{t \rightarrow \infty} \lim_{i \rightarrow \infty} R^{(t)}(x, C_i) = \lim_{i \rightarrow \infty} \xi[f^{-1}(C_i)] \\ = \xi[f^{-1}(\lim_{i \rightarrow \infty} C_i)] = \xi[f^{-1}(C)], \quad \forall x \notin N_1 \cup N_2.$$

This limit is also *uniform* with respect to (x, C) on $\mathbb{C}(N_1 \cup N_2) \times \mathcal{T}$, for, by (II.4.2), $t > n_\varepsilon$ and $x \notin N_1 \cup N_2$ implies that

$$-\varepsilon < R^{(t)}(x, C_i) - \xi[f^{-1}(C_i)] < \varepsilon, \quad \forall i \in \mathbf{N}^*,$$

and

$$-\varepsilon + \xi[f^{-1}(C)] < R^{(t)}(x, C) < \varepsilon + \xi[f^{-1}(C)].$$

Thus, $C \in \mathcal{T}'$, hence $\mathcal{T}' \supset \mathcal{T}$, consequently $\mathcal{T}' = \mathcal{T}$.

3° We examine now separately the three following cases:

(a) $s \geq 2$ and $y \in f(N_1 \cup N_2)$. Then $f^{-1}(y) \subset \mathbb{C}(N_1 \cup N_2)$ and, by (II.4.5), we have

$$(II.4.6) \quad \lim_{t \rightarrow \infty} Q_{s,t}(y, C) = \lim_{t \rightarrow \infty} R^{(t-s)}(x, C) = \xi[f^{-1}(C)], \quad x \in f^{-1}(y),$$

uniformly with respect to (y, C) on $\mathbb{C}(N_1 \cup N_2) \times \mathcal{T}$.

(b) $s \geq 2$ and $y \in f(N_1 \cup N_2)$. As indicated in (II.3.1), we take $Q_{s,s+1}(y, C) = \mu'(C)$ and, for $t > s+1$, we have

$$Q_{s,t}(y, C) = \int_{\mathcal{Y}} Q_{s,s+1}(y, dy') Q_{s+1,t}(y', C) \\ = \int_{\mathcal{Y}(N_1 \cup N_2)} Q_{s,s+1}(y, dy') Q_{s+1,t}(y', C).$$

By II.4.6 and the Fatou-Lebesgue theorem, we have

$$(II.4.7) \quad \lim_{t \rightarrow \infty} Q_{s,t}(y, C) = \int_{\mathcal{Y}(N_1 \cup N_2)} d\mu'(y') \lim_{t \rightarrow \infty} Q_{s+1,t}(y', C) = \xi[f^{-1}(C)],$$

uniformly with respect to (y, C) on $f(N_1 \cup N_2) \times \mathcal{T}$.

(c) $s = 1$. We have

$$\lim_{t \rightarrow \infty} Q_{1,t}(y, C) = \int_{\mathcal{Y}} Q_{1,2}(y, dy') \cdot \lim_{t \rightarrow \infty} Q_{2,t}(y', C) = \xi[f^{-1}(C)],$$

by (II.4.6) and (II.4.7), uniformly with respect to (y, C) on $\mathcal{Y} \times \mathcal{T}$.

Thus, the Markov process $((\mathcal{Y}, \mathcal{T}), Q_{t,t+1})_{t \in \mathbf{N}^*}$ is strongly and uniformly ergodic.

II.5. HYPOTHESES. Let $\mathcal{W}^* = \{y \in \mathcal{Y} : \mu[f^{-1}(y)] > 0\}$. We remark that \mathcal{W}^* contains an at most countable set of points, $\mathcal{W}^* = \{y_1, \dots, y_n, \dots\}$, for $\mu(\mathcal{X}) = 1$. I.3 (ii) implies that $\mathcal{W}^* \in \mathcal{T}$. We suppose that $\mu[f^{-1}(\mathcal{W}^*)] = 1$

and that $\forall x \in \mathcal{X}$, the series $\sum_k P_t(x, f^{-1}(y_k))$ converges uniformly with respect to t .

In the particular case where \mathcal{Y} is a finite or countable set, the condition $\mu[f^{-1}(\mathcal{W}^*)] = 1$ is always verified.

II.6. PROPOSITION. *If $(X_t)_{t \in \mathbf{N}^*}$ is attached to a strongly ergodic Markov process $((\mathcal{X}, \mathcal{B}), P_t)_{t \in \mathbf{N}^*}$ and if, moreover, the hypotheses II.2 and II.5 are satisfied, then $(f \circ X_t)_{t \in \mathbf{N}^*}$ is attached to a strongly ergodic Markov process.*

Proof. By hypothesis, there exists no μ -null set $N \in \mathcal{B}_f$ contained in $f^{-1}(\mathcal{W}^*)$. $\forall y_k \in \mathcal{W}^*$, there exists a μ -null set $N_k \in \mathcal{B}_f$ such that

$$\lim_{t \rightarrow \infty} R^{(t)}(x, \{y_k\}) = E_\mu[\lim_{t \rightarrow \infty} P_t(x, f^{-1}(y_k)) | \mathcal{B}_f] = \xi[f^{-1}(y_k)], \quad \forall x \notin N_k.$$

Hence, there exists $N = \bigcup_{k=1}^{\infty} N_k \in \mathcal{B}_f$, μ -null, such that

$$\lim_{t \rightarrow \infty} R^{(t)}(x, \{y_k\}) = \xi[f^{-1}(y_k)], \quad \forall x \notin N.$$

We must now prove that, $\forall C \subset \mathcal{W}^*$,

$$\lim_{t \rightarrow \infty} R^{(t)}(x, C) = \xi[f^{-1}(C)], \quad \forall x \notin N.$$

This formula is true for every C which is a finite union of $\{y_k\}$. It remains for us to examine the case where C is a countable union of $\{y_k\}$. Let

$C = \bigcup_{j=1}^{\infty} \{y_{k_j}\}$ be such a countable union. We have, $\forall x \notin N$,

$$\begin{aligned} \lim_{t \rightarrow \infty} R^{(t)}(x, C) &= \lim_{t \rightarrow \infty} \sum_{j=1}^{\infty} R^{(t)}(x, \{y_{k_j}\}) \\ &= \lim_{t \rightarrow \infty} \sum_{j=1}^{\infty} P_t(x, f^{-1}(y_{k_j})) \\ &= \sum_{j=1}^{\infty} \lim_{t \rightarrow \infty} P_t(x, f^{-1}(y_{k_j})) \\ &= \sum_{j=1}^{\infty} \lim_{t \rightarrow \infty} R^{(t)}(x, \{y_{k_j}\}) \\ &= \sum_{j=1}^{\infty} \xi[f^{-1}(y_{k_j})] \\ &= \xi[f^{-1}(C)]. \end{aligned}$$

For the rest of the proof, we use the same argument as that in II.4.3° with the three cases: $s \geq 2$ and $y \in \mathbb{C}(N)$; $s \geq 2$ and $y \in f(N)$; $s \geq 1$.

II.7. PROPOSITION. *If $(X_t)_{t \in \mathbb{N}^*}$ is attached to a weakly ergodic homogeneous Markov process, and if II.2 and II.5 are satisfied, then $(f \circ X_t)_{t \in \mathbb{N}^*}$ is attached to a weakly ergodic Markov process $((\mathcal{Y}, \mathcal{F}), Q_{t,t+1})_{t \in \mathbb{N}^*}$.*

Proof. We use, for the proof of this proposition, the following criteria for weak ergodicity (cf. [1]):

A Markov process $((A, \mathcal{L}), q_{t,t+1})_{t \in \mathbb{N}^*}$ is weakly ergodic if and only if for every $s \in \mathbb{N}^*$, $B \in \mathcal{L}$, for every increasing sequence $(t_j)_{j \in \mathbb{N}^*}$ of indices such that $q_{s,t_j}(u, B)$ converges to a limit when $j \rightarrow \infty$, for some $u \in A$, $q_{s,t_j}(v, B)$ converges $\forall v \in A$ to the same limit; moreover, this common limit is independent of s .

1° Let $(x_0, C) \in \mathcal{X} \times \mathcal{F}$, and let $\sigma = (t_i)_{i \in \mathbb{N}^*}$ be an increasing sequence of indices such that the sequence $(P_{t_i}(x_0, f^{-1}(C)))_{i \in \mathbb{N}^*}$ is convergent. The criteria cited above implies that $\forall s \in \mathbb{N}^*$ and $\forall x \in \mathcal{X}$, the sequence $(P_{t_i-s}(x, f^{-1}(C)))_{i \in \mathbb{N}^*}$ converges to the same limit. Let $l(\sigma, C)$ denote the common limit of these sequences.

An argument analogous to that in II.6.1° shows that

$$\lim_{i \rightarrow \infty} R^{(t_i-s)}(\cdot, C) = l(\sigma, C), \quad \mu\text{-a.s.},$$

and since there exists no μ -null element of \mathcal{B}_f , contained in $f^{-1}(\mathcal{B}^*)$, we have

$$\lim_{i \rightarrow \infty} R^{(t_i-s)}(x, C) = l(\sigma, C), \quad \forall x \in f^{-1}(\mathcal{B}^*).$$

Using the same argument employed in II.4.3°, with the 3 cases (a), (b), (c), we have $\forall C \in \mathcal{F}$, $\forall \sigma$,

$$\lim_{i \rightarrow \infty} Q_{s,t_i}(y, C) = l(\sigma, C), \quad \forall s \in \mathbb{N}^* \text{ and } \forall y \in \mathcal{Y}.$$

2° Now let $s_0 \in \mathbb{N}^*$, $C \in \mathcal{F}$ and let $\sigma = (t_j)_{j \in \mathbb{N}^*}$ be a sequence of indices such that $(Q_{s_0,t_j}(y_0, C))_{j \in \mathbb{N}^*}$ is convergent for a certain $y_0 \in \mathcal{Y}$. Let $l(\sigma, s_0, y_0, C)$ denote this limit. Following the criteria, we must prove that this limit is independent of y_0 and s_0 . Let us examine separately the 3 following cases:

(a) $s \geq 2$ and $y_0 \in \mathcal{B}^*$. Then, we have

$$Q_{s_0,t_j}(y_0, C) = R^{(t_j-s_0)}(x_0, C), \quad x_0 \in f^{-1}(y_0),$$

and

$$(II.7.1) \quad R^{(t_j-s_0)}(x_0, C) \cdot \mu[f^{-1}(y_0)] = \int_{f^{-1}(y_0)} d\mu(x) \cdot P_{t_j-s_0}(x, f^{-1}(C)).$$

The sequence $(P_{t_j-s_0}(x_0, f^{-1}(C)))_{j \in \mathbb{N}^*}$ being a sequence of numbers contained in the interval $[0, 1]$, we can find a convergent subsequence $(P_{t_{j_k}-s_0}(x_0, f^{-1}(C)))_{k \in \mathbb{N}^*}$. Since $((\mathcal{X}, \mathcal{B}), P_t)_{t \in \mathbb{N}^*}$ is weakly ergodic, the

limit of this convergent subsequence is independent of x_0 and of s_0 . Let $l(\sigma', C)$ denote this limit, σ' being the subsequence $(t_{j_k})_{k \in \mathbb{N}^*}$.

By the Fatou-Lebesgue theorem, (II.7.1) implies that

$$\lim_{j \rightarrow \infty} R^{(t_j-s_0)}(x_0, C) \cdot \mu[f^{-1}(y_0)] = \int_{f^{-1}(y_0)} d\bar{\mu}(x) \cdot \lim_{k \rightarrow \infty} P_{t_{j_k}-s_0}(x, f^{-1}(C)),$$

hence

$$l(\sigma, s_0, y_0, C) \cdot \mu[f^{-1}(y_0)] = l(\sigma', C) \cdot \mu[f^{-1}(y_0)].$$

Consequently, since $\mu[f^{-1}(y_0)] > 0$,

$$l(\sigma, s_0, y_0, C) = l(\sigma', C).$$

Thus, all the convergent subsequences of the sequence $(P_{t_j-s_0}(x_0, f^{-1}(C)))_{j \in \mathbb{N}^*}$ have the same limit: the sequence itself is then convergent, and

$$\lim_{j \rightarrow \infty} P_{t_j-s_0}(x_0, f^{-1}(C)) = l(\sigma, s_0, y_0, C).$$

By 1° $\forall s \in \mathbb{N}^*$ and $\forall y \in \mathcal{Y}$:

$$\lim_{j \rightarrow \infty} Q_{s,t_j}(y, C) = \lim_{j \rightarrow \infty} Q_{s_0,t_j}(y_0, C).$$

(b) $s_0 \geq 2$ and $y_0 \notin \mathcal{B}^*$. Then, $f^{-1}(y_0) \subset \mathbb{C}f^{-1}(\mathcal{B}^*)$. We have

$$Q_{s_0,t_j}(y_0, C) = \int_{\mathcal{B}^*} d\mu'(y) Q_{s_0+1,t_j}(y, C).$$

Let $y_1 \in \mathcal{B}^*$, and let $(Q_{s_0+1,t_j}(y_1, C))_{j \in \mathbb{N}^*}$ be a convergent subsequence. By (a), $\forall y \in \mathcal{Y}$, the sequence $((Q_{s_0+1,t_j}(y, C)))_{j \in \mathbb{N}^*}$ converges to the same limit, the Fatou-Lebesgue theorem gives

$$\lim_{k \rightarrow \infty} Q_{s_0,t_{j_k}}(y_0, C) = \int_{\mathcal{B}^*} d\mu'(y) \cdot \lim_{k \rightarrow \infty} Q_{s_0+1,t_{j_k}}(y, C),$$

that is to say

$$l(\sigma, s_0, y_0, C) = \lim_{k \rightarrow \infty} Q_{s_0+1,t_{j_k}}(y_1, C).$$

Thus, all the convergent subsequences of the sequence $(Q_{s_0+1,t_j}(y_0, C))_{j \in \mathbb{N}^*}$ have the same limit. Hence, the sequence itself is convergent, and

$$\lim_{j \rightarrow \infty} Q_{s_0+1,t_j}(y_1, C) = l(\sigma, s_0, y_0, C).$$

By 1° $\forall s \in \mathbb{N}^*$ and $\forall y \in \mathcal{Y}$,

$$\lim_{j \rightarrow \infty} Q_{s,t_j}(y, C) = \lim_{j \rightarrow \infty} Q_{s,t_j}(y_0, C).$$

(c) $s_0 = 1$. We have

$$Q_{1,t_j}(y_0, C) = \int_{\mathcal{Y}} Q_{1,2}(y_0, dy) Q_{2,t_j}(y, C).$$

Let $y_2 \in \mathcal{Y}$ and let $(Q_{2,t_k}(y_2, C))_{k \in \mathbb{N}^*}$ be a convergent subsequence. By (a) and (b), the sequence $(Q_{2,t_{j_k}}(y, C))_{k \in \mathbb{N}^*}$ converges to a limit independent of $y \in \mathcal{Y}$. The Fatou-Lebesgue theorem shows that

$$\lim_{k \rightarrow \infty} Q_{1,t_{j_k}}(y_0, C) = \int_{\mathcal{Y}} Q_{1,2}(y_0, dy) \lim_{k \rightarrow \infty} Q_{2,t_{j_k}}(y, C),$$

viz.

$$l(\sigma, s_0, y_0, C) = \lim_{k \rightarrow \infty} Q_{2,t_{j_k}}(y_2, C).$$

All the convergent subsequences of $(Q_{2,t_j}(y_2, C))_{j \in \mathbb{N}^*}$ having the same limit, the sequence itself is convergent, and we have

$$\lim_{j \rightarrow \infty} Q_{2,t_j}(y_2, C) = l(\sigma, s_0, y_0, C).$$

Consequently,

$$\lim_{j \rightarrow \infty} Q_{1,t_j}(y, C) = l(\sigma, s_0, y_0, C) = \lim_{j \rightarrow \infty} Q_{1,t_j}(y_0, C).$$

Thus, in every case, $l(\sigma, s_0, y_0, C)$ is independent of s_0 and y_0 , and the Markov process $((\mathcal{Y}, \mathcal{F}), Q_{t,t+1})_{t \in \mathbb{N}^*}$ is weakly ergodic.

II.8. EXAMPLES. Here are some examples in the case of a finite state space.

1° $\mathcal{X} = \{1, 2, 3, 4\}$, $\mathcal{B} = \mathcal{P}(\mathcal{X})$ and

$$P = \begin{pmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{6} & \frac{1}{6} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{6} & \frac{1}{6} & \frac{3}{3} & \frac{3}{3} \\ \frac{2}{10} & \frac{2}{10} & \frac{3}{10} & \frac{3}{10} \end{pmatrix}.$$

We can verify that the process $((\mathcal{X}, \mathcal{P}(\mathcal{X})), P_t)_{t \in \mathbb{N}^*}$ is strongly and uniformly ergodic, for \mathcal{X} is finite and

$$\lim_{t \rightarrow \infty} P_t = \begin{pmatrix} \frac{11}{47} & \frac{11}{47} & \frac{25}{94} & \frac{25}{94} \\ \frac{11}{47} & \frac{11}{47} & \frac{25}{94} & \frac{25}{94} \\ \frac{11}{47} & \frac{11}{47} & \frac{25}{94} & \frac{25}{94} \\ \frac{11}{47} & \frac{11}{47} & \frac{25}{94} & \frac{25}{94} \end{pmatrix}.$$

Now, consider $f: \mathcal{X} \rightarrow \mathcal{Y}$ such that $f(1) = f(2) = 1$; $f(3) = f(4) = 2$. We take μ on \mathcal{B}_f such that

$$\mu[f^{-1}(i)] = \sum_{j \in \mathcal{X}} a_j P(j, f^{-1}(i)), \quad \forall i \in \mathcal{Y} \text{ where } a_j > 0, \sum_{j \in \mathcal{X}} a_j = 1.$$

Hypothesis II.2, in our case, is satisfied by the mapping $R: \mathcal{X} \times \mathcal{P}(\mathcal{Y}) \rightarrow [0, 1]$ defined by

$$\begin{aligned} R(j_0, \{1\}) &= \frac{7}{12}, & R(j_0, \{2\}) &= \frac{5}{12} & \text{for } j_0 = 1, 2, \\ R(j_0, \{1\}) &= \frac{11}{30}, & R(j_0, \{2\}) &= \frac{19}{30} & \text{for } j_0 = 3, 4. \end{aligned}$$

Then, we have

$$Q_{t,t+1} = \begin{pmatrix} \frac{7}{12} & \frac{5}{12} \\ \frac{11}{30} & \frac{19}{30} \end{pmatrix}, \quad \forall t \geq 2.$$

Computations show that

$$\lim_{t \rightarrow \infty} Q_{s,t} = \begin{pmatrix} \frac{22}{47} & \frac{25}{47} \\ \frac{11}{30} & \frac{19}{30} \end{pmatrix}, \quad \forall s \geq 2,$$

and

$$\lim_{t \rightarrow \infty} Q_{1,t} = Q_{1,2} \cdot \lim_{t \rightarrow \infty} Q_{2,t} = \lim_{t \rightarrow \infty} Q_{2,t},$$

for $Q_{1,2}$ is a stochastic matrix and $\lim_{t \rightarrow \infty} Q_{2,t}$ is with all its rows identical.

Thus, $((\mathcal{Y}, \mathcal{P}(\mathcal{Y})), Q_{t,t+1})_{t \in \mathbb{N}^*}$ is also strongly and uniformly ergodic. 2° $\mathcal{X} = \{1, 2, 3\}$, $\mathcal{B} = \mathcal{P}(\mathcal{X})$ and

$$P = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix}.$$

The Markov process $((\mathcal{X}, \mathcal{P}(\mathcal{X})), P_t)_{t \in \mathbb{N}^*}$ is strongly and uniformly ergodic, for \mathcal{X} is finite and $P_t = P, \forall t \in \mathbb{N}^*$. Let $\mathcal{Y} = \{1, 2\}$ and $f: \mathcal{X} \rightarrow \mathcal{Y}$ be defined by $f(1) = f(2) = 1$, and $f(3) = 2$.

We take μ on \mathcal{B}_f such that

$$\mu[f^{-1}(i)] = \sum_{j \in \mathcal{X}} a_j P(j, f^{-1}(i)), \quad \forall i \in \mathcal{Y} \text{ where } a_j > 0, \sum_{j \in \mathcal{X}} a_j = 1.$$

Thus,

$$\mu[f^{-1}(1)] = 1 \quad \text{and} \quad \mu[f^{-1}(2)] = 0.$$

Hypothesis II.2 is satisfied by the mapping $R: \mathcal{X} \times \mathcal{P}(\mathcal{Y}) \rightarrow [0, 1]$ defined by

$$\begin{aligned} R(1, \{1\}) &= R(2, \{1\}) = 1, \\ R(1, \{2\}) &= R(2, \{2\}) = 0, \\ R(3, \{1\}) + R(3, \{2\}) &= 1, \end{aligned}$$

where $R(3, \{1\})$ can be arbitrarily chosen.

Hence, we can take

$$Q_{t,t+1} = \begin{pmatrix} 1 & 0 \\ \nu[f^{-1}(1)] & \nu[f^{-1}(2)] \end{pmatrix} \quad \text{for } t \geq 2,$$

where ν is an arbitrarily chosen probability measure on \mathcal{B}_f .

a) If ν is such that $\nu[f^{-1}(1)] = 0$ and $\nu[f^{-1}(2)] = 1$, then

$$Q_{t,t+1} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \quad \forall t \geq 2,$$

and the process $((\mathcal{B}, \mathcal{P}(\mathcal{B})), Q_{t,t+1})_{t \in \mathbb{N}^*}$ is not strongly ergodic.

b) If ν is such that $\nu[f^{-1}(1)] = \alpha$, where $\alpha \in]0, 1[$, then

$$Q_{t,t+1} = \begin{pmatrix} 1 & 0 \\ \alpha & 1-\alpha \end{pmatrix}, \quad \forall t \geq 2.$$

We have

$$\lim_{t \rightarrow \infty} Q_{s,s+t} = \lim_{t \rightarrow \infty} \begin{pmatrix} 1 & 0 \\ 1-(1-\alpha)^t & (1-\alpha)^t \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}, \quad \forall s \geq 2,$$

and, consequently, the process $((\mathcal{B}, \mathcal{P}(\mathcal{B})), Q_{t,t+1})_{t \in \mathbb{N}^*}$ is strongly ergodic.

We examine now the converse for the proposition II.7.

II.9. HYPOTHESES. (i) $\forall x \in \mathcal{X}$, $P(x, \cdot)$ is of the form

$$(II.9.1) \quad P(x, B) = \int_B g(x, x') d\bar{\mu}(x'),$$

where $B \in \mathcal{B}$, and $g(x, \cdot)$ is \mathcal{B}_f -measurable.

(ii) $\forall (x, B) \in \mathcal{X} \times \mathcal{B}$, if $(t_j)_{j \in \mathbb{N}^*}$ is an increasing sequence of indices such that the sequence $(P_{t_j}(x, B))_{j \in \mathbb{N}^*}$ converges, then $\forall x' \in f^{-1}(x)$, the sequence $(P_{t_j}(x', B))_{j \in \mathbb{N}^*}$ converges and

$$(II.9.2) \quad \lim_{j \rightarrow \infty} P_{t_j}(x, B) = \lim_{j \rightarrow \infty} P_{t_j}(x', B).$$

II.10. The following is an interpretation of II.9: II.9 (i) implies the condition of Doeblin (cf. [2]) for the homogeneous Markov process $((\mathcal{X}, \mathcal{B}), P_t)_{t \in \mathbb{N}^*}$. II.9 (ii) implies that $\forall y \in \mathcal{Y}$, if $f^{-1}(y)$ is included in an ergodic set (cf. [2]) decomposable in cyclic subsets $\Gamma_1, \dots, \Gamma_d$, then $f^{-1}(y)$ is included in one of these cyclic subsets. Indeed, suppose that there exist two distinct cyclic subsets Γ_k and $\Gamma_{k'}$ such that $\Gamma_k \cap f^{-1}(y) \neq \emptyset$ and $\Gamma_{k'} \cap f^{-1}(y) \neq \emptyset$. We know, by [2], that there exist probabilities ${}_k\pi$ and ${}_{k'}\pi$ on \mathcal{B} such that ${}_k\pi(\Gamma_k) = 1$ and ${}_{k'}\pi(\Gamma_{k'}) = 1$ and that

$$\lim_{n \rightarrow \infty} P_{nd}(x_k, \Gamma_k) = {}_k\pi(\Gamma_k) = 1,$$

$$\lim_{n \rightarrow \infty} P_{nd}(x_{k'}, \Gamma_{k'}) = {}_{k'}\pi(\Gamma_{k'}) = 0,$$

and thus, II.9 (ii) is not verified.

II.11. PROPOSITION. Under II.9 (i), $\forall t \in \mathbb{N}^*$ and $\forall (x, B) \in \mathcal{X} \times \mathcal{B}$,

$$P_t(x, B) = \int_B g^{(t)}(x, x') d\bar{\mu}(x'),$$

where the $g^{(t)}$, \mathcal{B}_f -measurable, are defined by recurrence:

$$g^{(1)} = g,$$

$$g^{(t)}(x, x') = \int_{\mathcal{X}} g^{(t-1)}(x, x_1) g(x_1, x') d\bar{\mu}(x_1), \quad \forall x \text{ and } x' \in \mathcal{X}.$$

Proof. We proceed by induction. It is easy to verify the formula for $t = 2$. Suppose now that

$$P_{t-1}(x, B) = \int_B g^{(t-1)}(x, x') d\bar{\mu}(x').$$

We have

$$\begin{aligned} P_t(x, B) &= \int_{\mathcal{X}} P_{t-1}(x, dx_1) P(x_1, B) \\ &= \int_{\mathcal{X}} g^{(t-1)}(x, x_1) d\bar{\mu}(x_1) \cdot \int_B g(x_1, x') d\bar{\mu}(x') \\ &= \int_B \left[\int_{\mathcal{X}} g^{(t-1)}(x, x_1) g(x_1, x') d\bar{\mu}(x_1) \right] d\bar{\mu}(x') \\ &= \int_B g^{(t)}(x, x') d\bar{\mu}(x'). \end{aligned}$$

II.12. PROPOSITION. Suppose that II.5 and II.9 are satisfied. Let $((\mathcal{Y}, \mathcal{T}), Q_{t,t+1})_{t \in \mathbb{N}^*}$, to which $(f \circ X_t)_{t \in \mathbb{N}}$ is attached, be defined by (II.3.1) with $N = f^{-1}(\mathcal{Y}^*)$. If $((\mathcal{Y}, \mathcal{T}), Q_{t,t+1})_{t \in \mathbb{N}^*}$ is strongly ergodic, then $((\mathcal{X}, \mathcal{B}), P_t)_{t \in \mathbb{N}^*}$ is also strongly ergodic.

Proof. By hypothesis, $\forall s \in \mathbb{N}^*$, $\forall y \in \mathcal{Y}$ and $\forall C \in \mathcal{T}$,

$$(II.12.1) \quad \lim_{t \rightarrow \infty} Q_{s,s+t}(y, C) = \pi(C), \quad \text{where } \pi \text{ is a probability on } \mathcal{T}.$$

1° At first, let us examine the limit $\lim_{t \rightarrow \infty} P_t(x, B)$, for $x \in \mathcal{X}$ and $B \in \mathcal{B}_f$.

Let $x_0 \in \mathcal{X}$, $y_0 = f(x_0)$ and $C \in \mathcal{T}$.

(a) If $x_0 \in f^{-1}(\mathcal{Y}^*)$, then $f^{-1}(y_0) = f^{-1}(\mathcal{Y}^*)$, and

$$Q_{s,s+t}(y_0, C) = R^{(t)}(x_0, C).$$

Hence,

$$(II.12.2) \quad Q_{s,s+t}(y_0, C) \cdot \mu[f^{-1}(y_0)] = \int_{f^{-1}(y_0)} d\bar{\mu}(x) P_t(x, f^{-1}(C)).$$

$(P_t(x_0, f^{-1}(C)))_{t \in \mathbb{N}^*}$ being a sequence of numbers of $[0, 1]$, we can find a convergent subsequence $(P_{t_j}(x_0, f^{-1}(C)))_{j \in \mathbb{N}^*}$ whose limit is $l(x_0, \sigma, C)$,

σ denoting the increasing sequence $(t_j)_{j \in \mathbb{N}^*}$ of indices. By II.9 (ii), we have $\forall x \in f^{-1}(y_0)$,

$$(II.12.3) \quad \lim_{j \rightarrow \infty} P_{t_j}(x, f^{-1}(C)) = l(x_0, \sigma, C).$$

Consequently, the Fatou-Lebesgue theorem, (II.12.1), (II.12.2) and (II.12.3) imply that

$$\lim_{j \rightarrow \infty} Q_{s, s+t_j}(y_0, C) \cdot \mu[f^{-1}(y_0)] = \int_{f^{-1}(y_0)} \bar{\mu}(x) \cdot \lim_{j \rightarrow \infty} P_{t_j}(x, f^{-1}(C)),$$

viz.

$$\lim_{j \rightarrow \infty} Q_{s, s+t_j}(y_0, C) \cdot \mu[f^{-1}(y_0)] = l(x_0, \sigma, C) \cdot \mu[f^{-1}(y_0)],$$

viz.

$$\pi(C) = l(x_0, \sigma, C), \quad \text{for } \mu[f^{-1}(y_0)] > 0.$$

All the convergent subsequences of the sequence $(P_t(x_0, f^{-1}(C)))_{t \in \mathbb{N}^*}$ having the same limit, the sequence itself is then convergent and we have

$$(II.12.4) \quad \lim_{t \rightarrow \infty} P_t(x_0, f^{-1}(C)) = \pi(C), \quad \forall x_0 \in f^{-1}(\mathscr{U}^*) \text{ and } \forall C \in \mathscr{T}.$$

(b) If $x_0 \in \mathbb{E} f^{-1}(\mathscr{U}^*)$, we have, for $t \geq 2$,

$$P_t(x_0, f^{-1}(C)) = \int_{\mathbb{E}} P(x_0, dx) P_{t-1}(x, f^{-1}(C)).$$

Since $\mu[f^{-1}(\mathscr{U}^*)] = 1$, we have $P(x, f^{-1}(\mathscr{U}^*)) = 1, \forall x \in \mathbb{E}$. Hence,

$$P_t(x_0, f^{-1}(C)) = \int_{f^{-1}(\mathscr{U}^*)} P(x_0, dx) P_{t-1}(x, f^{-1}(C)),$$

so that, by (II.12.4) and the Fatou-Lebesgue theorem,

$$(II.12.5) \quad \lim_{t \rightarrow \infty} P_t(x_0, f^{-1}(C)) = \int_{f^{-1}(\mathscr{U}^*)} P(x_0, dx) \lim_{t \rightarrow \infty} P_{t-1}(x, f^{-1}(C)) = \pi(C).$$

Thus, (a) and (b) show that $\forall (y, C) \in \mathbb{E} \times \mathscr{T}$,

$$(II.12.6) \quad \lim_{t \rightarrow \infty} P_t(y, f^{-1}(C)) = \pi(C).$$

2° We examine now the limit $\lim_{t \rightarrow \infty} P_t(x, B)$ for $x \in \mathbb{E}$ and $B \in \mathscr{B}$. Let x and $x' \in \mathbb{E}$, and $y' = f(x')$. By (II.12.6) and II.10, we have

$$\lim_{t \rightarrow \infty} P_t(x, f^{-1}(y')) = \pi(\{y'\}) = \lim_{t \rightarrow \infty} \int_{f^{-1}(y')} g^{(t)}(x, x_1) \bar{\mu}(x_1).$$

$g^{(t)}(x, \cdot)$ being by hypothesis \mathscr{B}_f -measurable, is then constant on $f^{-1}(y')$ which is an atom of \mathscr{B}_f . Hence,

$$\pi(\{y'\}) = \lim_{t \rightarrow \infty} g^{(t)}(x, x') \cdot \mu[f^{-1}(y')].$$

We then conclude that, $\forall x \in \mathbb{E}$, the sequence $(g^{(t)}(x, \cdot))_{t \in \mathbb{N}^*}$ converges μ -a.s. to a limit g' independent of $x \in \mathbb{E}$.

Consequently,

$$(II.12.7) \quad \begin{aligned} \lim_{t \rightarrow \infty} P_t(x, B) &= \lim_{t \rightarrow \infty} \int_B g^{(t)}(x, x_1) \bar{\mu}(x_1) \\ &= \int_B \lim_{t \rightarrow \infty} g^{(t)}(x, x_1) \bar{\mu}(x_1) = \int_B g'(x_1) \bar{\mu}(x_1). \end{aligned}$$

This last equality proves the strong ergodicity of $((\mathbb{E}, \mathscr{B}), P_t)_{t \in \mathbb{N}^*}$.

In fact, we have a stronger result than Proposition II.12, but its proof makes use of the latter.

II.13. PROPOSITION. Suppose that II.5 and II.9 are satisfied. Let $((\mathscr{U}, \mathscr{T}), Q_{t, t+1})_{t \in \mathbb{N}^*}$, to which $(f \circ X_t)_{t \in \mathbb{N}^*}$ is attached, be defined by (II.3.1) with $N = f^{-1}(\mathscr{U}^*)$. If $((\mathscr{U}, \mathscr{T}), Q_{t, t+1})_{t \in \mathbb{N}^*}$ is weakly ergodic, then $((\mathbb{E}, \mathscr{B}), P_t)_{t \in \mathbb{N}^*}$ is strongly ergodic.

Proof. 1° II.9 (i) implies that the Markov process $((\mathbb{E}, \mathscr{B}), P_t)_{t \in \mathbb{N}^*}$ verifies the Doeblin condition. The Markov process $((\mathscr{U}, \mathscr{T}), Q_{t, t+1})_{t \geq 2}$ defined by (II.3.1) with $N = f^{-1}(\mathscr{U}^*)$ is homogeneous. It also verifies the Doeblin condition, for $s \geq 2$.

2° We know (cf. [1]) that, for a homogeneous Markov process satisfying the Doeblin condition, weak ergodicity and strong ergodicity are equivalent. Consequently, the Markov process $((\mathscr{U}, \mathscr{T}), Q_{t, t+1})_{t \geq 2}$ is strongly ergodic. Then, $\forall s \geq 2$,

$$\forall (y, C) \in \mathscr{U} \times \mathscr{T}, \quad \lim_{t \rightarrow \infty} Q_{s, s+t}(y, C) = \pi(C).$$

3° Let us prove now that $((\mathscr{U}, \mathscr{T}), Q_{t, t+1})_{t \in \mathbb{N}^*}$ is strongly ergodic. Indeed, for $s = 1$, and $\forall (y, C) \in \mathscr{U} \times \mathscr{T}$,

$$\begin{aligned} \lim_{t \rightarrow \infty} Q_{1, 1+t}(y, C) &= \lim_{t \rightarrow \infty} \int_{\mathscr{U}} Q_{1, 2}(y, dy') Q_{2, 1+t}(y', C) \\ &= \int_{\mathscr{U}} Q_{1, 2}(y, dy') \lim_{t \rightarrow \infty} Q_{2, 1+t}(y', C) = \pi(C). \end{aligned}$$

4° By III.12, we see that the process $((\mathbb{E}, \mathscr{B}), P_t)_{t \in \mathbb{N}^*}$ is strongly ergodic.

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On the zeroes of some random functions

by

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Let $F(t)$ be a Fourier series with random coefficients and phases,

$$F(t) = \sum_{n=1}^{\infty} a_n X_n \cos(nt + \Phi_n).$$

Here $(X_n)_{n=1}^{\infty}$ is a sequence of mutually independent Gaussian variables of type $N(0, 1)$; $(\Phi(n))_{n=1}^{\infty}$ is a sequence of mutually independent variables, uniformly distributed upon $[0, 2\pi]$; and the X 's and Φ 's are mutually independent. (The basic probability space will be denoted (Ω, P) .) About the numbers a_n we suppose

$$a_n > 0, \quad \log a_n = -\beta \log n + o(\log n), \quad \text{with } \frac{1}{2} < \beta \leq 1.$$

Our goal is an estimation of the zero-set of F , $Z(\omega) = \{t: F(t, \omega) = 0\}$.

THEOREM. Let B be a closed set in $[0, 2\pi]$ of Hausdorff dimension d .

Then

$$P\{\dim(Z \cap B) \leq d - \beta + \frac{1}{2}\} = 1, \quad d \geq \beta - \frac{1}{2},$$

$$P\{Z \cap B = \emptyset\} = 1, \quad d < \beta - \frac{1}{2},$$

$$P\{\dim(Z \cap B) \geq d_1\} > 0, \quad 0 < d_1 < d - \beta + \frac{1}{2}.$$

In § 1 we prove a general principle for the lower bound, whose application is dependent upon specific estimates, derived later about F . In § 2 we review some conclusions from [2] about the uniform convergence of F and its modulus of continuity, and we also obtain a technical lemma about the local character of the trajectories of F . In § 3 we obtain an upper bound for the dimension, and in § 4 a lower bound is obtained by combining the work of §§ 1 and 2.

§ 1. Let B be a compact set of real numbers and μ a probability measure in B such that:

(i) $\mu([a, a+h]) \leq C_1 h^d$ for constants $C_1, d > 0$ and all intervals $[a, a+h]$.