

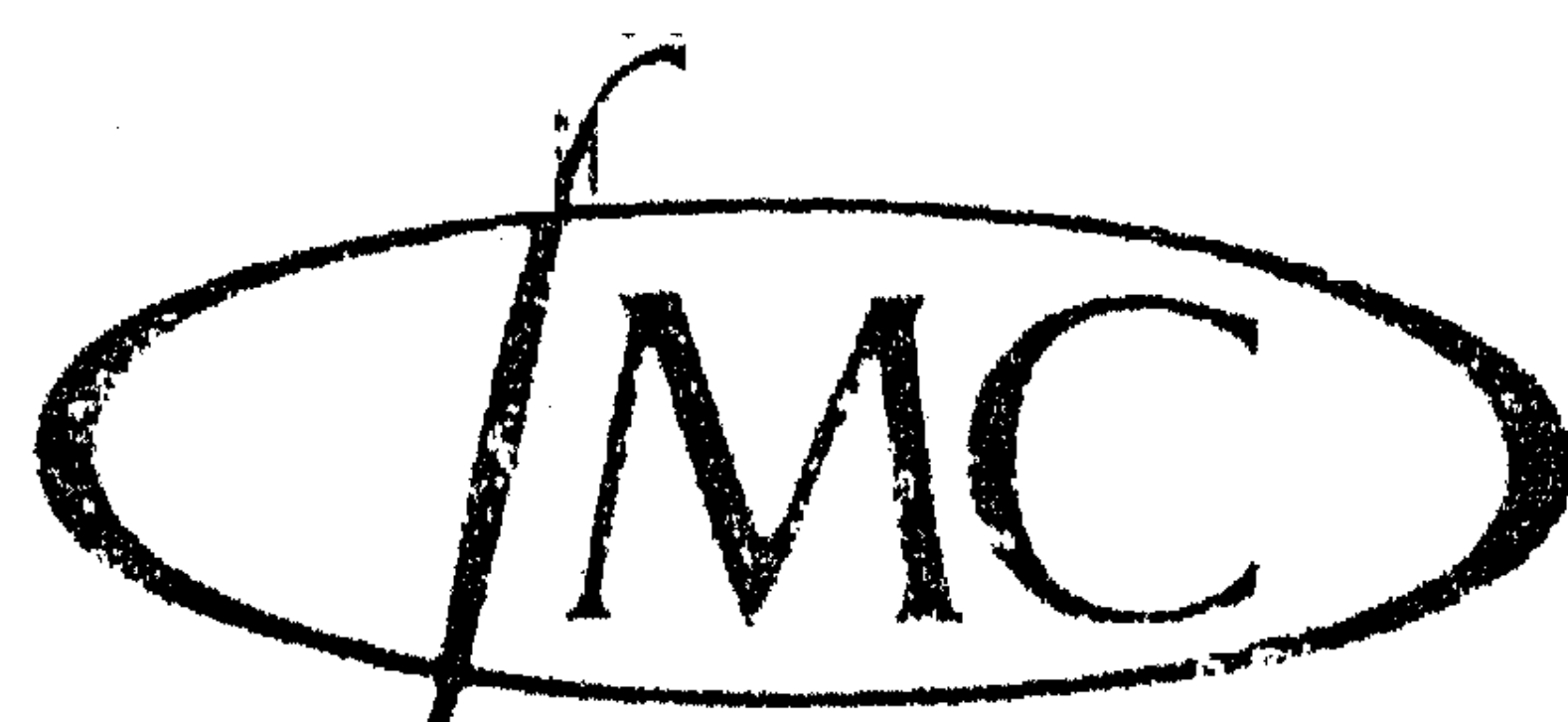
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On stable transformations

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ON STABLE TRANSFORMATIONS

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Summary. Let T be a measure preserving transformation of a probability space (Ω, \mathcal{A}, P) into itself.

We will say that T is a stable transformation if for every

$A, B \in \mathcal{A}$, $\lim_{n \rightarrow \infty} P(T^{-n} A \cap B)$ exists.

Stable transformations are investigated in this article with the aid of Rényi's results on stable sequences of events. The concept of a stable transformation generalises that of a mixing transformation.

1. Introduction

Let (Ω, \mathcal{A}, P) be a probability space.

Let T be a measurable transformation (not necessarily one to one) of Ω into itself. Assume further that T is measure preserving, that is, $P(T^{-1} A) = P(A)$ for every $A \in \mathcal{A}$. Following Rényi [5], we will say that T is stable if for every $A \in \mathcal{A}$, $\{T^{-n} A, n = 1, 2, \dots\}$ is a stable sequence of sets, that is, for every $A, B \in \mathcal{A}$, $\lim_{n \rightarrow \infty} P(T^{-n} A \cap B)$ exists. The purpose of this article is to study such transformations.

The concept of stability generalises that of mixing. It will be shown that a stable transformation T is mixing if and only if the σ -field of invariant sets is trivial. [A measurable set A is said to be invariant if $T^{-1} A = A$].

As the present investigation relies heavily on the results proved in [5], we will for the sake of completeness give a résumé of these in section 2. In section 3 the analogues of results for stable sequences of sets will be proved for stable transformations. Examples of stable transformations will be given in section 4.

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2. Resumé of results on stable sequences of events

Let (Ω, \mathcal{A}, P) be a probability space and let $\{A_n, n = 1, 2, \dots\}$ be a sequence of events. We will say that $\{A_n\}$ is a stable sequence of events if for every $B \in \mathcal{A}$

$$\lim_{n \rightarrow \infty} P(A_n \cap B) = Q(B)$$

exists.

Theorem 2.1. If $\{A_n\}$ is a stable sequence of events and Q is as above, then Q is a measure on (Ω, \mathcal{A}) and is absolutely continuous with respect to P .

Denote by α the Radon-Nikodym derivative of Q with respect to P . α is said to be the local density of the stable sequence of sets $\{A_n\}$.

A sequence of events $\{A_n, n = 1, 2, \dots\}$ is said to be mixing if there exists $\beta, 0 \leq \beta \leq 1$ such that for every $B \in \mathcal{A}$

$$\lim_{n \rightarrow \infty} P(A_n \cap B) = \beta P(B).$$

β is called the density of the mixing sequence $\{A_n\}$.

Corollary 2.1. If $\{A_n\}$ is stable with local density α , then $\{A_n\}$ is mixing if and only if α is a constant almost surely.

Theorem 2.2. The sequence of events $\{A_n, n = 1, 2, \dots\}$ is stable if and only if

$$\lim_{n \rightarrow \infty} P(A_k \cap A_n) = Q_k, \quad k = 1, 2, \dots$$

exists. If, in addition, $P(A_k) > 0, k = 1, 2, \dots$, set $q_k = \frac{Q_k}{P(A_k)}$, $k = 1, 2, \dots$, and $q_0 = \lim_{n \rightarrow \infty} P(A_n)$. Then $\{A_n\}$ is mixing if and only if the q_k 's ($k = 0, 1, 2, \dots$) are all equal.

The property of stability is preserved if the underlying probability measure P is replaced by a probability measure absolutely continuous with respect to P . More explicitly, we have

Theorem 2.3. Let $\{A_n, n = 1, 2, \dots\}$ be a stable sequence of events with local density α on the probability space (Ω, \mathcal{A}, P) . Let P^* be a probability measure on (Ω, \mathcal{A}) , absolutely continuous with respect to P . Then $\{A_n\}$ is stable on $(\Omega, \mathcal{A}, P^*)$ with local density α .

3. Some general theorems on stable transformations

Let T be a stable transformation on (Ω, \mathcal{A}, P) , that is, T is measure preserving and $\lim_{n \rightarrow \infty} P(T^{-n} A \cap B)$ exists for every $A, B \in \mathcal{A}$. The limit is easy to find.

Theorem 3.1. Let T be a stable transformation. Then

$$\lim_{n \rightarrow \infty} P(T^{-n} A \cap B) = \int_B P(A/\mathcal{J}) dP$$

for every $A, B \in \mathcal{A}$. Here \mathcal{J} is the invariant σ -field and $P(A/\mathcal{J})$ is the conditional probability of A given \mathcal{J} .

Proof. By definition, the sequence $\{T^{-n} A, n = 1, 2, \dots\}$, where $A \in \mathcal{A}$, is stable. Hence $\lim_{n \rightarrow \infty} P(T^{-n} A \cap B)$ exists for every $B \in \mathcal{A}$. But by the Individual Ergodic Theorem, we have:

$\frac{1}{n} \sum_{k=0}^{n-1} I_{T^{-k} A}$ converges almost surely to $P(A/\mathcal{J})$, where I_C is the indicator of the set C . Hence if $B \in \mathcal{A}$, $\frac{1}{n} \sum_{k=0}^{n-1} I_{T^{-k} A} \cdot I_B$ converges almost surely to $P(A/\mathcal{J}) \cdot I_B$. Apply the Dominated Convergence Theorem. We get:

$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} P(T^{-k} A \cap B) = \int_B P(A/\mathcal{J}) dP$, that is, the sequence $\{P(T^{-n} A \cap B)\}$ is Cesaro-summable to $\int_B P(A/\mathcal{J}) dP$. The result now follows from the remark made at the beginning of the proof.

Remark. Denote by α_A the local density of the stable sequence $\{T^{-n} A\}$, $A \in \mathcal{A}$. What we have proved then is that $\int_B \alpha_A dP = \int_B P(A/\mathcal{J}) dP$ for every $B \in \mathcal{A}$. But α_A and $P(A/\mathcal{J})$ are \mathcal{A} -measurable functions. Hence $\alpha_A = P(A/\mathcal{J})$ almost surely. Therefore the local density of $\{T^{-n} A\}$ is simply $P(A/\mathcal{J})$.

In order to check if a measure preserving transformation T is stable, it is in fact sufficient to verify that $\lim_{n \rightarrow \infty} P(T^{-n} A \cap B)$ exists for $A = B \in \mathcal{A}$.

Theorem 3.2. A measure preserving transformation T is stable if and only if $\lim_{n \rightarrow \infty} P(T^{-n} A \cap A)$ exists for every $A \in \mathcal{A}$.

Proof. The "only if" part is trivial. Consider now the sequence $\{T^{-n} A, n = 1, 2, \dots\}$, $A \in \mathcal{A}$. We want to show that $\{T^{-n} A\}$ is stable. Note that since T is measure preserving, $P(T^{-k} A \cap T^{-n} A) = P(T^{-k}(T^{-(n-k)} A \cap A)) =$

$= P(T^{-(n-k)} A \cap A)$, where $n > k$. But by the hypothesis, $\lim_{n \rightarrow \infty} P(T^{-(n-k)} A \cap A)$ exists and so $\lim_{n \rightarrow \infty} P(T^{-k} A \cap T^{-n} A)$ exists, $k = 1, 2, \dots$. Hence, by Theorem 2.2., $\{T^{-n} A\}_{n=1}^{\infty}$ is stable. This completes the "if" part of the proof.

A measure preserving transformation T is mixing if for every $A \in \mathcal{A}$, the sequence of events $\{T^{-n} A, n = 1, 2, \dots\}$ is mixing with density $P(A)$, that is, if for every $A, B \in \mathcal{A}$

$$\lim_{n \rightarrow \infty} P(T^{-n} A \cap B) = P(A) \cdot P(B).$$

Clearly a mixing transformation is stable. When is the converse true?

Corollary 3.1. In order that a stable transformation T be mixing, it is necessary and sufficient that \mathcal{I} , the σ -field of invariant sets, be trivial under P .

Proof. Suppose that \mathcal{I} is trivial under P , that is, if $A \in \mathcal{I}$ then $P(A) = 0$ or 1 . By Theorem 3.1., since T is stable, we have

$$\lim_{n \rightarrow \infty} P(T^{-n} A \cap B) = \int_B P(A/\mathcal{I}) dP$$

for every $A, B \in \mathcal{A}$. But as \mathcal{I} is trivial, $P(A/\mathcal{I}) = P(A)$ almost surely for every $A \in \mathcal{A}$. Hence $\lim_{n \rightarrow \infty} P(T^{-n} A \cap B) = P(A) \cdot P(B)$ for every $A, B \in \mathcal{A}$, so that T is mixing. Conversely, assume that T is mixing. Let $A \in \mathcal{I}$. Then $T^{-n} A = A$ for $n = 1, 2, \dots$. But $\{T^{-n} A, n = 1, 2, \dots\}$ is mixing. Hence for every $B \in \mathcal{A}$, $P(A \cap B) = P(A) \cdot P(B)$, that is, $P(A) = 0$ or 1 . Therefore, \mathcal{I} is trivial, which concludes the proof.

Let us now turn to the functional form of stability. Let $\mathcal{L}_2(\Omega, \mathcal{A}, P)$ be the class of complex-valued random variables f on (Ω, \mathcal{A}, P) such that $\int |f|^2 dP < \infty$. Identify all functions in \mathcal{L}_2 which differ on a set of measure zero. Then \mathcal{L}_2 is a Hilbert space over the field of complex numbers with inner-product $(f, g) = \int f \bar{g} dP$ (here \bar{x} is the complex-conjugate of x) and norm $\|f\| = (\int |f|^2 dP)^{1/2}$. If T is a measure preserving transformation of Ω into itself we can define a transformation U of \mathcal{L}_2 into itself as follows: $Uf = f \circ T$, $f \in \mathcal{L}_2$. Then U is an isometry, that is, U is a bounded linear transformation such that $\|Uf\| = \|f\|$ for every $f \in \mathcal{L}_2$ (see [2], page 14). Denote by U^n the n -th iterate of U .

Call a function $f \in \mathcal{L}_2$ invariant if $Uf = f$. Denote by E_0 the projection

on the closed subspace of invariant functions in \mathcal{L}_2 . We can now characterize stability of T as follows.

Theorem 3.3. A measure preserving transformation T is stable if and only if $\lim_{n \rightarrow \infty} (U^n f, g) = (E_0 f, g)$ for every $f, g \in \mathcal{L}_2$ that is, U^n converges to E_0 in the weak operator topology.

Proof. The proof depends on the remark that the conditional expectation of f given \mathcal{J} is almost surely equal to $E_0 f$. If f and g are indicators of sets F and G respectively, then the functional form simply reduces to the set-theoretic definition of stability. To go the other way, use a double approximation process as follows: let g be a fixed indicator in \mathcal{L}_2 . The result holds for simple functions $f \in \mathcal{L}_2$ and so by \mathcal{L}_2 -approximation holds for functions $f \in \mathcal{L}_2$. Now let f be a fixed function in \mathcal{L}_2 and a similar argument about g yields the result.

In the case of mixing, \mathcal{J} is trivial so that all invariant functions in \mathcal{L}_2 are constants. Hence $E_0 f = (f, 1)1$ for every $f \in \mathcal{L}_2$, where 1 stands for the function which is equal to one everywhere.

Corollary 3.2. A measure preserving transformation T is mixing if and only if $\lim_{n \rightarrow \infty} (U^n f, g) = ((f, 1)1, g) = (f, 1)(1, g)$ for every $f, g \in \mathcal{L}_2$.

We may add here that if T is stable, then U^n converges to E_0 in the strong operator topology only in a rather trivial and uninteresting case. In fact, U^n converges to E_0 in the strong operator topology if and only if every function in \mathcal{L}_2 is invariant. To prove this statement, note that since U^n converges weakly to E_0 , U^n will converge strongly to E_0 if and only if $\lim_{n \rightarrow \infty} \|U^n f\| = \|E_0 f\|$ for each $f \in \mathcal{L}_2$. But $\|U^n f\| = \|f\|$. Note also that for any $f \in \mathcal{L}_2$, $\|f\|^2 = \|E_0 f\|^2 + \|f - E_0 f\|^2$ by the Decomposition Theorem. Hence $\|E_0 f\| = \|f\|$ for each $f \in \mathcal{L}_2$ if and only if $E_0 f = f$ for each $f \in \mathcal{L}_2$. This completes the proof.

The property of stability is preserved if the underlying measure is replaced by a measure absolutely continuous with respect to it. Explicitly we have:

Theorem 3.4. Let T be a stable transformation on (Ω, \mathcal{A}, P) . Let Q be a probability measure on (Ω, \mathcal{A}) such that Q is absolutely continuous with respect to P . Assume further that Q is preserved by T . Then T is stable

on (Ω, \mathcal{A}, Q) and for every $A \in \mathcal{A}$, $P(A/g) = Q(A/g)$ almost surely $[Q]$.

Proof. Consider the sequence of sets $\{T^{-n} A, n = 1, 2, \dots\}$, $A \in \mathcal{A}$. Since Q is absolutely continuous with respect to P , by Theorem 2.3., $\{T^{-n} A\}$ is stable with respect to Q . Hence T is stable on (Ω, \mathcal{A}, Q) . Furthermore, by Theorem 2.3., $\lim_{n \rightarrow \infty} Q(T^{-n} A \cap B) = \int_B P(A/g) dQ$ for every $A, B \in \mathcal{A}$. Hence by Theorem 3.1 we have: $\int_B Q(A/g) dQ = \int_B P(A/g) dQ$ for every $A, B \in \mathcal{A}$. This proves the second assertion of the theorem.

Corollary 3.3. Let P and Q be probability measures on (Ω, \mathcal{A}) . Assume that T is stable for both P and Q . Then, if $P = Q$ on \mathcal{G} , $P = Q$ on \mathcal{A} .

Proof. Let $\mu(A) = \frac{1}{2} P(A) + \frac{1}{2} Q(A)$, $A \in \mathcal{A}$. It is easy to verify that T is stable for μ . Note that P, Q are absolutely continuous with respect to μ . Furthermore, $\mu = P = Q$ on \mathcal{G} . By Theorem 3.4, $\mu(A/g) = P(A/g)$ almost surely $[P]$ for every $A \in \mathcal{A}$. Note that the exceptional set above is \mathcal{G} -measurable and so must have μ -measure zero as well. Again, as $P(A/g), \mu(A/g)$ are \mathcal{G} -measurable functions, we have:

$$\mu(A) = \int \mu(A/g) d\mu^g = \int P(A/g) dP^g = P(A)$$

for every $A \in \mathcal{A}$. Here μ^g, P^g denote the restrictions of μ, P , respectively to \mathcal{G} . This proves the corollary.

Corollary 3.4. Let T be a mixing transformation on (Ω, \mathcal{A}, P) . Let Q be a probability measure on (Ω, \mathcal{A}) . Assume that Q is absolutely continuous with respect to P and that it is preserved by T . Then $P = Q$.

Proof. Follows directly from Theorem 3.4.

Corollary 3.5. Let P and Q be probability measures on (Ω, \mathcal{A}) for which T is a mixing transformation. Then either $P = Q$ or P and Q are mutually singular.

Proof. Suppose $P \neq Q$. Then by Corollary 3.3, there exists a set $A \in \mathcal{G}$ such that $P(A) \neq Q(A)$. But, since T is mixing for both P and Q , either $P(A) = 1$ and $Q(A) = 0$ or $P(A) = 0$ and $Q(A) = 1$. In either case, P and Q are mutually singular.

4. Examples of stable transformations

A. Let T be the identity transformation on a probability space (Ω, \mathcal{A}, P) , that is, $T\omega = \omega$, $\omega \in \Omega$. Then $\mathcal{I} = \mathcal{A}$ and T is stable. If \mathcal{A} is non-trivial, we get an example of a stable transformation that is not mixing.

B. Let (Ω, \mathcal{A}) be a countably infinite product of a measurable space $(\Omega_0, \mathcal{A}_0)$. Denote by ω_n ($n = 1, 2, \dots$) the n -th coordinate of a point ω in Ω . We will use the following notation for finite dimensional rectangles:
 $C(E_1^{(i_1)}, \dots, E_n^{(i_n)})$ is the set of all ω such that $\omega_{i_k} \in E_k$, $k = 1, \dots, n$.
 If $i_k = k$, $k = 1, \dots, n$, we will write $C(E_1, \dots, E_n)$. Let T be the shift operation on Ω , that is, $T\omega = \omega^1$, where $\omega_n^1 = \omega_{n+1}$, $n = 1, 2, \dots$.
 Consider a symmetric probability measure P on (Ω, \mathcal{A}) , that is

$$P(C(E_1^{(i_1)}, \dots, E_n^{(i_n)})) = P(C(E_1^{(j_1)}, \dots, E_n^{(j_n)}))$$

for all $n = 1, 2, \dots$, all $E_1, E_2, \dots, E_n \in \mathcal{A}_0$ and all sequences of positive integers i_1, \dots, i_n and j_1, \dots, j_n (i 's all distinct and j 's all distinct).

Then T is a stable transformation on (Ω, \mathcal{A}, P) . To see this, first note that T is measure preserving. Now let B be a measurable $\{1, \dots, m\}$ -cylinder, that is $B = A \times \Omega_0 \times \Omega_0 \times \dots$ where A is a measurable subset of $\Omega_0 \times \Omega_0 \times \dots \times \Omega_0$ (m times). Consider the sequence of sets $\{B_k, k = 1, 2, \dots\}$ where $B_k = T^{-k} B$, $k = 1, 2, \dots$. It is clear that B_k is a $\{k+1, \dots, k+m\}$ -cylinder with base B . Hence, as P is a symmetric measure, for n large $P(B_k \cap B_n) = P(C)$, where C is the $\{1, \dots, 2m\}$ -cylinder $B \times B \times \Omega_0 \times \Omega_0 \times \dots$. Hence $\lim_{n \rightarrow \infty} P(B_k \cap B_n)$ exists for $k = 1, 2, \dots$. Therefore, by Theorem 2.2,

$T^{-k} B^{n \rightarrow \infty}$ is stable. But for every set $A \in \mathcal{A}$ and $\epsilon > 0$, there exists a $\{1, \dots, m\}$ -cylinder B (for some m) such that $P(A \Delta B) < \epsilon$. It is easy to see that the stability of the sequence $\{T^{-n} A\}$ follows from that of $\{T^{-n} B\}$.

This proves that T is a stable transformation.

In particular, let P be a product measure with identical components. Then it is well known that T is mixing (see [6], page 110). Conversely, assume that T is mixing for a symmetric measure P . Let $A = C(E_1, \dots, E_m)$ be a measurable finite-dimensional rectangle. It is easy to see that

$$\lim_{n \rightarrow \infty} P(T^{-k} A \cap T^{-n} A) = P(C(E_1, \dots, E_m, E_1, \dots, E_m)), k = 1, 2, \dots$$

The limit is independent of k . But the sequence $\{T^{-n} A\}$ is mixing. Hence, by Theorem 2.2, we must have

$$P(C(E_1, \dots, E_m, E_1, \dots, E_m)) = P^2(C(E_1, \dots, E_m)).$$

As T is mixing, this last relation is true for all measurable finite-dimensional rectangles. Hence, by Theorems 5.2 and 5.3 in [3] (see pages 477-478), P must be a product measure with identical components. Hence we have

Theorem 4.1. Let P be a symmetric probability measure on (Ω, \mathcal{A}) . Then T is stable and T is mixing if and only if P is a product measure with identical components.

C. Let $\{x_n, n=0, 1, \dots\}$ be a stationary aperiodic Markov chain with countable state space I . Elements of I will be denoted by i with or without subscripts. Assume that the Markov chain is defined on the appropriate (unilateral) sequence space (Ω, \mathcal{A}) and let T be the shift operator on (Ω, \mathcal{A}) . If P is the relevant probability measure on (Ω, \mathcal{A}) , T is stable on (Ω, \mathcal{A}, P) .

To prove this, let us note that it is sufficient to demonstrate stability of sequences of events $\{T^{-n} A, n = 1, 2, \dots\}$, where A is a finite-dimensional rectangle of the form $(x_0 = i_0, \dots, x_m = i_m)$, the i 's being ergodic states belonging to the same class. We have for large n

$$P(T^{-k} A \cap T^{-n} A) = p_{i_0} p_{i_0 i_1} \dots p_{i_{m-1} i_m} p_{i_m i_0}^{(n-m-k)} p_{i_0 i_1} \dots p_{i_{m-1} i_m}$$

where p_i denotes the initial distribution, p_{ij} the one-step transition probability and $p_{ij}^{(n)}$ the n -step transition probability.

Clearly since $\lim_{n \rightarrow \infty} p_{ij}^{(n)} = \pi_{ij}$ for j ergodic,

$$\lim_{n \rightarrow \infty} P(T^{-k} A \cap T^{-n} A) = p_{i_0} p_{i_0 i_1} \dots p_{i_{m-1} i_m} \pi_{i_m i_0} p_{i_0 i_1} \dots p_{i_{m-1} i_m},$$

$k = 1, 2, \dots$

Hence by Theorem 2.2, $\{T^{-n} A\}$ is stable. This proves the assertion.

D. We conclude with an example of a measure preserving transformation which is not stable.

Let $\Omega = [0, 1]$, \mathcal{A} the σ -field of Borel subsets of Ω , P Lebesgue measure on \mathcal{A} . Let T be an invertible, both ways measurable, measure preserving transformation of Ω onto Ω , which has strict period m ($m > 1$) at almost all $[P]$ points of Ω .

According to a result of Halmos (see [2], page 70), there exists a set $E \in \mathcal{A}$ such that $P(E) = 1/m$ and $E, T^{-1}E, \dots, T^{-(m-1)}E$ are pairwise disjoint. It follows that $\limsup P(T^{-n}E \cap E) = 1/m$ and $\liminf P(T^{-n}E \cap E) = 0$, so that the sequence $\{T^{-n}E, n = 1, 2, \dots\}$ is not stable. Hence T is not stable.

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Résumé. Soit T une transformation, conservant la mesure, d'un espace de probabilité (Ω, \mathcal{A}, P) dans lui-même. On dira que T est stable si, pour tout $A, B \in \mathcal{A}$, il existe $\lim_{n \rightarrow \infty} P(T^{-n}A \cap B)$. L'investigation des transformations stables est fondée sur des résultats de Rényi concernant les suites stables d'événements. La notion de transformation stable est une généralisation de celle de transformation mélangée.