Hausdorff Dimension of Some Continued-Fraction Sets

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Section 1

Introduction. Each irrational ω in the unit interval Ω has a unique continued-fraction expansion

$$\omega = \frac{1}{a_1(\omega)} + \frac{1}{a_2(\omega)} + \cdots$$

with integral partial quotients $a_j(\omega)$. We shall compute the Hausdorff dimension of certain sets defined in terms of the frequencies with which the $a_j(\omega)$ assume various values.

Let $N_n(i, \omega)$ be the number of $j, 1 \le j \le n$, for which $a_j(\omega) = i$. For a probability vector $p = (p_1, p_2, ...)$, let L(p) be the set of ω for which

$$\lim_{n \to \infty} \frac{1}{n} N_n(i, \omega) = p_i, \quad i = 1, 2, ...,$$
(1.1)

and for a set A of probability vectors, let $L(A) = \bigcup_{p \in A} L(p)$. Thus L(A) is the set of ω for which the frequency vector $(n^{-1}N_n(1,\omega), n^{-1}N_n(2,\omega), \ldots)$ approaches (in each component) some element of A.

We first give a lower bound for dim L(A) in terms of entropy. Call a probability measure v on the Borel subsets of Ω stationary if under v the stochastic process $[a_1(\omega), a_2(\omega), ...]$ is stationary; in this case, let h(v) be the entropy of the process. Call v ergodic if under v the process is ergodic. Put

$$t(v) = -2 \int \log \omega v(d\omega). \tag{1.2}$$

On the set of k-tuples of positive integers, define a probability measure v_k by

$$v_k(i_1, \dots, i_k) = v[\omega: a_j(\omega) = i_j, j \leq k].$$

$$(1.3)$$

If for each k, P_k is a probability measure on the k-tuples, and if $\{P_k\}$ is consistent, then there exists a v such that $v_k = P_k$ for all k.

The entropy h(v) is finite if and only if $-\sum_i v_1(i) \log v_1(i)$ converges. The quantity t(v) is always positive, and since $a_1(\omega) \le \omega^{-1} \le a_1(\omega) + 1$, it is finite if and only if $\sum_i v_1(i) \log i$ converges. Let \mathcal{N} be the set of stationary and ergodic v for which h(v) and t(v) are both finite.

Theorem 1. For each set A of probability vectors,

$$\dim L(A) \ge \sup \left[\frac{h(v)}{t(v)} : v \in \mathcal{N}, v_1 \in A \right].$$
(1.4)

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This result, related to theorems of Kinney and Pitcher [6], is proved in Section 2. For inequalities going the other way, we must restrict attention to the set Ω_r of ω for which $a_j(\omega) \leq r, j = 1, 2, ...$ If A is a set of r-dimensional probability vectors $p = (p_1, ..., p_r)$, let $U_r(A)$ be the set of ω in Ω_r such that for every positive ε and n_0 there exist a p in A and an integer $n \geq n_0$ that satisfy

$$\left|\frac{1}{n}N_n(i,\omega) - p_i\right| < \varepsilon, \quad i = 1, \dots, r.$$
(1.5)

Thus $U_r(A)$ is the set of ω in Ω_r for which some limit point of the frequency vectors $(n^{-1}N_n(1,\omega), \ldots, n^{-1}N_n(r,\omega))$ lies in the closure of the set A.

Let \mathcal{N}_r be the set of v in \mathcal{N} supported by Ω_r . A v in \mathcal{N} lies in \mathcal{N}_r if and only if $v_1(i)=0$ for i>r; in this case we may regard v_1 as an *r*-dimensional probability vector. Finally, let \mathcal{M}_r consist of the v in \mathcal{N}_r under which $[a_1(\omega), a_2(\omega), ...]$ is a Markov chain of some order.

Theorem 2. For each set A of r-dimensional probability vectors,

dim
$$U_r(A) \leq \sup \left[\frac{h(v)}{t(v)}: v \in \mathcal{M}_r, v_1 \in A\right].$$
 (1.6)

This theorem is proved in Section 4. If p is an r-dimensional probability vector, let $L_r(p)$ be the set of ω in Ω_r that satisfy (1.1) for $i \leq r$, and put $L_r(A) = \bigcup_{p \in A} L_r(p)$. Since $L_r(A) \subset U_r(A)$ and $\mathcal{M}_r \subset \mathcal{N}_r$, it is a simple matter (see Section 3) to combine Theorems 1 and 2:

Theorem 3. For each set A of r-dimensional probability vectors,

$$\dim L_r(A) = \dim U_r(A) = \sup \left[\frac{h(v)}{t(v)} : v \in \mathcal{N}_r, v_1 \in A \right]$$

$$= \sup \left[\frac{h(v)}{t(v)} : v \in \mathcal{M}_r, v_1 \in A \right].$$
(1.7)

Let C_k be the set of r^k -dimensional vectors p with nonnegative components $p(i_1, \ldots, i_k)$ satisfying the constraint

$$\sum p(i_1, \dots, i_k) = 1 \tag{1.8}$$

(in this sum, as in (1.11), the indices range from 1 to r) and, for each i_1, \ldots, i_{k-1} , the constraint

$$\sum_{i} p(i_1, \dots, i_{k-1}, i) = \sum_{i} p(i, i_1, \dots, i_{k-1}).$$
(1.9)

Define $\tau_k: C_k \to C_1$ by

$$(\tau_k(p))_i = \sum_{i_1, \dots, i_{k-1}} p(i, i_1, \dots, i_{k-1}).$$
 (1.10)

Let $q_k(i_1, \ldots, i_k)$ be the denominator of the fraction

$$\frac{1}{i_1} + \dots + \frac{1}{i_k}$$

(in lowest terms). Finally, for $A \subset C_1$, define

$$\alpha_k(A) = \sup_{\substack{p \in C_k \\ \tau_k(p) \in A}} \frac{-\sum p(i_1, \dots, i_k) \log p(i_1, \dots, i_k)}{2\sum p(i_1, \dots, i_k) \log q_k(i_1, \dots, i_k)}$$
(1.11)

(with the convention $0 \log 0 = 0$). In all this, r is fixed and hence does not appear in the notation.

Theorem 4. For each $A \subset C_1$,

$$\dim L_r(A) = \dim U_r(A) = \lim_{k \to \infty} \alpha_k(A). \tag{1.12}$$

This theorem is proved in Section 5. Since $U_r(C_1) = \Omega_r$, (1.7) and (1.12) determine dim Ω_r . Good [5] (see also [7, 8]) has given a different expression for dim Ω_r ; Section 6 deals with the connection between the two results and with a conjecture the connection suggests.

The sets $L_r(A)$ and $U_r(A)$ have analogues defined in terms of the frequencies with which the v-tuples $(a_j(\omega), \ldots, a_{j+v-1}(\omega))$ assume various values. It will be clear from the proofs that the corresponding analogues of the above theorems hold, with the appropriate restrictions on v_v in the suprema. It will also be clear that the set $[1, \ldots, r]$ to which the partial quotients have been restricted can be replaced by any finite set.

Section 2

Proof of Theorem 1. For each n, Ω splits into the fundamental intervals of order n, the various sets

$$[\omega: a_j(\omega) = i_j, j \leq n];$$

let $u_n(\omega)$ be that fundamental interval of order *n* containing ω . If *v* is stationary and ergodic and the entropy h(v) is finite, then according to [4],

$$\lim_{n \to \infty} \left[-\frac{1}{n} \log v(u_n(\omega)) \right] = h(v)$$
(2.1)

except on a set of v-measure 0.

In all that follows, λ will represent Lebesgue measure. If $q_n(\omega)$ is the denominator of the *n*th convergent, then $\lambda^{-1}(u_n(\omega)) = q_n(\omega)(q_n(\omega) + q_{n-1}(\omega))$ (see [3; (4.10)], for example), and hence

$$2\log q_n(\omega) \leq -\log \lambda(u_n(\omega)) \leq 2\log q_n(\omega) + \log 2.$$
(2.2)

Moreover [3; (4.21)],

$$\left|\log q_n(\omega) + \sum_{j=1}^n \log \left(\frac{1}{a_j(\omega)} + \frac{1}{a_{j+1}(\omega)} + \cdots \right) \right| \leq 4.$$
(2.3)

If v is stationary and ergodic and l(v) is finite, then

$$-\frac{2}{n}\sum_{j=1}^{n}\log\left(\frac{1}{|a_{j}(\omega)|}+\frac{1}{|a_{j+1}(\omega)|}+\cdots\right) \to t(v)$$

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almost everywhere by the ergodic theorem, and hence, by (2.2) and (2.3),

$$\lim_{n\to\infty} \left[-\frac{1}{n} \log \lambda(u_n(\omega)) \right] = t(v).$$
(2.4)

If $v \in \mathcal{N}$, then (2.1) and (2.4) both hold except for ω in a set of v-measure 0.

Let dim_v M denote dimension relative to v and to coverings by fundamental intervals. That is, let $v_{\alpha}(M, \rho) = \inf \sum_{i} v(v_i)^{\alpha}$, the infimum extending over countable coverings $[v_i]$ of M by fundamental intervals with $v(v_i) < \rho$, let

$$v_{\alpha}(M) = \lim_{\rho \to 0} v_{\alpha}(M, \rho),$$

and take dim_v M as that α_0 such that $v_{\alpha}(M) = \infty$ for $\alpha < \alpha_0$ and $v_{\alpha}(M) = 0$ for $\alpha > \alpha_0$. See [1].

By Theorem 2.4 of [2] (or see [3; Section 14]), if $\log v(u_n(\omega))/\log \lambda(u_n(\omega)) \rightarrow \delta$ on M, then $\dim_{\lambda} M = \delta \dim_{\nu} M$, so that $\dim_{\lambda} M = \delta$ if v(M) > 0. By Lemma 3.1 of [6], if $-n^{-1}\log \lambda(u_n(\omega))$ goes to a positive constant on M, then $\dim_{\lambda} M$ coincides with ordinary Hausdorff dimension dim M. Thus we have the following result.

Lemma 1. If $\alpha \ge 0$ and $\beta > 0$, then the three relations

$$M \subset \left[\omega: \lim_{n \to \infty} \left[-\frac{1}{n} \log v(u_n(\omega)) \right] = \alpha \right],$$
(2.5)

$$M \subset \left[\omega : \lim_{n \to \infty} \left[-\frac{1}{n} \log \lambda(u_n(\omega)) \right] = \beta \right],$$
(2.6)

$$v(M) > 0,$$
 (2.7)

together imply

$$\dim M = \frac{\alpha}{\beta}.$$
 (2.8)

Let M_v be the intersection of the two sets on the right in (2.5) and (2.6) with $\alpha = h(v)$ and $\beta = t(v)$. By (2.1) and (2.4), $v \in \mathcal{N}$ implies $v(M_v) = 1$. Applying Lemma 1 to $M \cap M_v$, we see that v(M) > 0 implies dim $M \ge \dim M \cap M_v = h(v)/t(v)$, which gives this result:

Lemma 2. For arbitrary M,

$$\dim M \ge \sup \left[\frac{h(v)}{t(v)} : v \in \mathcal{N}, v(M) > 0 \right].$$
(2.9)

If $v \in \mathcal{N}$, and $v_1 = p = (p_1, p_2, ...)$, then by the ergodic theorem, v(L(p)) = 1. Therefore, by Lemma 2,

dim
$$L(p) \ge \sup\left[\frac{h(v)}{t(v)}: v \in \mathcal{N}, v_1 = p\right],$$
 (2.10)

and Theorem 1 follows from this and the definition of L(A).

Section 3

Proof of Theorem 3. If $v \in \mathcal{N}_r$, then $v(\Omega_r) = 1$, so that, in addition to (2.10), Lemma 2 gives

dim
$$L_r(p) \ge \sup \left[\frac{h(v)}{t(v)}: v \in \mathcal{N}_r, v_1 = p\right]$$

for r-dimensional probability vectors p, and hence

$$\dim L_r(A) \ge \sup \left[\frac{h(v)}{t(v)}; v \in \mathcal{N}_r, v_1 \in A \right]$$
(3.1)

for sets A of such vectors. Since (1.7) is a consequence of (1.6), (3.1), and the relations $L_r(A) \subset U_r(A)$ and $\mathcal{M}_r \subset \mathcal{N}_r$, Theorem 3 will follow if we prove Theorem 2.

Section 4

Proof of Theorem 2. Since dim $M \leq \dim_{\lambda} M$, the following result is a consequence of Theorem 2.1 of [2].

Lemma 3. For $\delta \geq 0$,

$$\dim \left[\omega \colon \liminf_{n \to \infty} \frac{-n^{-1} \log v(u_n(\omega))}{-n^{-1} \log \lambda(u_n(\omega))} \leq \delta \right] \leq \delta.$$
(4.1)

The proof of (1.6) is based on this lemma. To find an upper bound for the limit inferior in (4.1), we require an upper bound for the numerator of the ratio and a lower bound for the denominator; these we shall find separately.

We first show that

$$t(v) > \frac{1}{6} \tag{4.2}$$

and

$$-2\int \log\left(\frac{1}{|a_1(\omega)|} + \dots + \frac{1}{|a_k(\omega)|}\right) v(d\omega) > \frac{1}{6}, \quad k \ge 3,$$

$$(4.3)$$

for stationary v. If $v_1(1) = p$, then $v_3(1, 1, 1) \ge 1 - 3(1-p)$. The quantities in question are at least

$$-2(1-v_1(1))\log \frac{1}{2} = 2(1-p)\log 2,$$

which exceeds $\frac{1}{6}$ for $0 \le p \le \frac{3}{4}$; they are also at least

$$-2v_{3}(1, 1, 1)\log\left(\frac{1}{1} + \frac{1}{1} + \frac{1}{1}\right) \ge 2(3p-2)\log\frac{3}{2},$$

which exceeds $\frac{1}{6}$ for $\frac{3}{4} \leq p \leq 1$.

From now on, r is a fixed positive integer, each ω considered lies in Ω_r , and the indices i_1, i_2, \ldots range from 1 to r. We shall bound the numerator in (4.1) for probability measures v corresponding to higher-order Markov chains. As in Section 1, let C_k be the set of r^k -dimensional vectors p with nonnegative components satisfying (1.8) and (1.9). Let C_k^0 be the set of p in C_k with strictly positive components.

For p in C_k put

$$s_p(i_1, \dots, i_{k-1}) = \sum_i p(i_1, \dots, i_{k-1}, i)$$
 (4.4)

and define

$$t_{p}(i_{1}, \dots, i_{k-1}; i_{k}) = \begin{cases} \frac{p(i_{1}, \dots, i_{k})}{s_{p}(i_{1}, \dots, i_{k-1})} & \text{if } s_{p}(i_{1}, \dots, i_{k-1}) > 0\\ \delta_{i_{k-1}, i_{k}} & \text{otherwise.} \end{cases}$$
(4.5)

Now (4.5) is a set of transition probabilites for a Markov chain of order k-1 with states 1, ..., r. Because of (1.8) and (1.9), (4.4) are stationary probabilites for the chain. There is a probability measure on Ω_r under which $[a_1(\omega), a_2(\omega), ...]$ is a stationary chain with these stationary and transition probabilities; call this measure v_p .

For p in C_k , define (here and later \sum extends over all k-tuples from [1, ..., r])

$$h_k(p) = -\sum p(i_1, \dots, i_k) \log t_p(i_1, \dots, i_{k-1}; i_k)$$
(4.6)

(with $0 \log 0 = 0$). The entropy $h(v_p)$ of v_p is just $h_k(p)$. Clearly $h_k(p)$ is continuous on C_k^0 ; that it is continuous on all of C_k follows from the fact that

$$-p(i_1,\ldots,i_k)\log t_p(i_1,\ldots,i_{k-1};i_k)$$

vanishes for $p(i_1, \ldots, i_k) = 0$, and is at most $-p(i_1, \ldots, i_k) \log p(i_1, \ldots, i_k)$ for positive $p(i_1, \ldots, i_k)$ and hence tends to 0 with $p(i_1, \ldots, i_k)$. Also define

$$I_k(p) = -2\sum p(i_1, \dots, i_k) \log\left(\frac{1}{i_1} + \dots + \frac{1}{i_k}\right).$$

$$(4.7)$$

Clearly $t_k(p)$ is continuous on C_k .

Let $N_n(i_1, \ldots, i_k; \omega)$ be the number of $j, 1 \leq j \leq n$, for which

$$(a_j(\omega), \dots, a_{j+k-1}(\omega)) = (i_1, \dots, i_k).$$
 (4.8)

If $p \in C_k^0$, then

$$-\log v_p(u_n(\omega)) = -\sum N_{n-k+1}(i_1, \dots, i_k; \omega) \log t_p(i_1, \dots, i_{k-1}; i_k) -\log s_p(a_1(\omega), \dots, a_{k-1}(\omega)),$$
(4.9)

a formula useful for bounding the numerator in (4.1). Let $N'_n(i_1, \ldots, i_k; \omega)$ be the number of $j, 1 \le j \le n-k+1$, for which (4.8) holds, plus the number of $j, n-k+2 \le j \le n$, for which

$$(a_{j}(\omega), \dots, a_{n}(\omega), a_{1}(\omega), \dots, a_{j-n-1+k}(\omega)) = (i_{1}, \dots, i_{k})$$
(4.10)

holds. Let $\pi_n(k, \omega)$ have components $\pi_n(i_1, \ldots, i_k; \omega) = n^{-1} N_n(i_1, \ldots, i_k; \omega)$, and let $\pi'_n(k, \omega)$ have components $\pi'_n(i_1, \ldots, i_k; \omega) = n^{-1} N'_n(i_1, \ldots, i_k; \omega)$. Then $\pi'_n(k, \omega)$ satisfies (1.9) as well as (1.8) and hence lies in C_k (which is not generally true of $\pi_n(k, \omega)$). Clearly $N_{n-k+1} \leq N'_n$, so (4.9) implies

$$-\log v_p(u_n(\omega)) \leq -\sum \pi'_n(i_1, \dots, i_k; \omega) \log t_p(i_1, \dots, i_{k-1}; i_k) -\frac{1}{n} \log s_p(a_1(\omega), \dots, a_{k-1}(\omega)).$$
(4.11)

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For p in C_k^0 and $\varepsilon > 0$, let V_p^{ε} be the set of π in C_k for which

$$-\sum |\pi(i_1, \dots, i_k) - p(i_1, \dots, i_k)| \log t_p(i_1, \dots, i_{k-1}; i_k) < \varepsilon$$
(4.12)

and

$$-2\sum |\pi(i_1,\ldots,i_k)-p(i_1,\ldots,i_k)|\log\left(\frac{1}{i_1}+\cdots+\frac{1}{i_k}\right)<\varepsilon.$$
(4.13)

Note that V_p^{ε} is open relative to C_k . If $p \in C_k^0$, if $\pi'_n(k, \omega) \in V_p^{\varepsilon}$, and if *n* exceeds some $n_1(\varepsilon, p, \omega)$, then by (4.11),

$$-\frac{1}{n}\log v_p(u_n(\omega)) \leq h_k(p) + 2\varepsilon.$$
(4.14)

To estimate the corresponding denominator in (4.1), note first that

$$-\log \lambda(u_n(\omega)) \ge -2\sum_{j=1}^n \log \left(\frac{1}{a_j(\omega)} + \frac{1}{a_{j+1}(\omega)} + \cdots \right) - 8$$

by (2.2) and (2.3). Since

$$\left|\log\left(\frac{1}{b_1} + \frac{1}{b_2} + \cdots\right) - \log\left(\frac{1}{b_1} + \cdots + \frac{1}{b_k}\right)\right| \leq \frac{4}{2^k}$$
(4.15)

(see [3; (4.7)], for example), we have

$$-\log \lambda(u_n(\omega)) \ge -2\sum N_n(i_1,\ldots,i_k;\omega) \log \left(\frac{1}{i_1} + \cdots + \frac{1}{i_k}\right) - 8 - \frac{8n}{2^k}$$

Since $i_1 \leq r$, and since N_n and N'_n differ by less than k, this gives

$$-\frac{1}{n}\log\lambda(u_n(\omega)) \ge -2\sum \pi'_n(i_1,\ldots,i_k;\omega)\log\left(\left\lceil \frac{1}{i_1} + \cdots + \frac{1}{i_k}\right)\right)$$
$$-\frac{2k}{n}r^k\log(r+1) - \frac{8}{n} - \frac{8}{2^k}.$$

Therefore, if $p \in C_k^0$, if $\pi'_n(k, \omega) \in V_p^{\varepsilon}$, and if *n* exceeds some $n_2(\varepsilon, p)$, then

$$-\frac{1}{n}\log\lambda(u_n(\omega)) \ge t_k(p) - 2\varepsilon - 8 \cdot 2^{-k}.$$
(4.16)

From now on consider only $k \ge 7$ and $\varepsilon < 1/24$; for such pairs $2\varepsilon + 8 \cdot 2^{-k} < \frac{1}{6}$, and so the right side of (4.16) is positive by (4.3). Suppose p lies in C_k^0 . If $\pi_n(k, \omega)$ lies in V_p^{ε} i.o. (that is, for infinitely many values of n), then $\pi'_n(k, \omega)$ lies in $V_p^{2\varepsilon}$ i.o., so that (4.14) and (4.16) with 2ε in place of ε are simultaneously true i.o. It follows by Lemma 3 that

$$\dim \left[\omega \in \Omega_r : \pi_n(k, \omega) \in V_p^{\varepsilon} \text{ i.o.} \right] \leq \frac{h_k(p) + 4\varepsilon}{I_k(p) - 4\varepsilon - 8 \cdot 2^{-k}}.$$
(4.17)

For a nonempty subset B of C_k , define

$$G(B) = \left[\omega \in \Omega_r \colon \liminf_{n \to \infty} d(\pi_n(k, \omega), B) = 0\right], \tag{4.18}$$

where d denotes ordinary distance in r^k -space. Let $B^{\varepsilon} = [\pi \in C_k: d(\pi, B) < \varepsilon]$. If $\pi \in C_k$ and $\varepsilon > 0$, then there exists a p in C_k^0 such that $d(\pi, p) < \varepsilon$ and $\pi \in V_p^{\varepsilon}$ (because the sums in (4.12) and (4.13) go to 0 as p approaches π and because C_k is convex and C_k^0 is its interior relative to the hyperplane determined by (1.8) and (1.9)). Thus the V_p^{ε} with p in $B^{\varepsilon} \cap C_k^0$ cover the closure of B, and therefore B can be covered by finitely many $V_{p_1}^{\varepsilon}, \ldots, V_{p_a}^{\varepsilon}$ with $p_a \in B^{\varepsilon} \cap C_k^0$. Clearly

$$G(B) \subset \bigcup_{\alpha=1}^{a} \left[\omega \in \Omega_r \colon \pi_n(k, \omega) \in V_{p_\alpha}^{\varepsilon} \text{ i.o.} \right].$$

Therefore dim G(B) is at most the supremum of the right side of (4.17) with p ranging over $B^{\varepsilon} \cap C_{k}^{0}$. Since ε is arbitrary, and since (4.6) and (4.7) are continuous in p,

dim
$$G(B) \leq \sup_{p \in B} \frac{h_k(p)}{t_k(p) - 8 \cdot 2^{-k}}$$
. (4.19)

Now suppose that A is a nonempty set of r-dimensional probability vectors – a subset of C_1 . With τ_k defined by (1.10), $G(\tau_k^{-1}A)$ coincides with $U_r(A)$ as defined in Section 1 (see (1.5)). Now (4.19) applied to $\tau_k^{-1}A$ gives

dim
$$U_r(A) \leq \sup_{p \in \tau_k^{-1}A} \frac{h_k(p)}{t_k(p) - 8 \cdot 2^{-k}}.$$
 (4.20)

For p in C_k , it follows by (1.9) that

$$\sum_{i} s_{p}(i, i_{1}, \dots, i_{k-2}) = \sum_{i} s_{p}(i_{1}, \dots, i_{k-2}, i).$$

Call this common value $c_p(i_1, \ldots, i_{k-2})$ and define

$$p^{*}(i_{1}, \dots, i_{k}) = \frac{s_{p}(i_{1}, \dots, i_{k-1}) s_{p}(i_{2}, \dots, i_{k})}{c_{p}(i_{2}, \dots, i_{k-1})}$$
(4.21)

if the denominator is positive, and define $p^*(i_1, \ldots, i_k) = 0$ otherwise. Then p^* lies in C_k and has the same stationary probabilities as p; moreover $t_{p^*}(i_1, \ldots, i_{k-1}; i_k) > 0$ if $s_p(i_1, \ldots, i_{k-1})$ and $s_p(i_2, \ldots, i_k)$ are both positive. Any convex combination $xp + (1-x)p^*, 0 \le x < 1$, has the same properties and the corresponding Markov chain is therefore ergodic. Thus any p in C_k can be approximated by elements π of C_k for which v_{π} is ergodic and for which the stationary probabilities are as for p. Since (4.6) and (4.7) are continuous in p, the supremum in (4.20) is unaltered if in addition to $p \in \tau_k^{-1} A$ we require that v_p be ergodic.

Now $t_k(p)$ differs from $t(v_p)$ by at most $8 \cdot 2^{-k}$ because of (4.15), and $h_k(p)$ coincides with $h(v_p)$. Therefore (if $k \ge 7$, so that $16 \cdot 2^{-k} < \frac{1}{6}$)

dim
$$U_r(A) \leq \sup \frac{h(v)}{t(v) - 16 \cdot 2^{-k}}$$
,

the supremum extending over all ergodic Markov chains of order k-1 with $v_1 \in A$. Since a chain of order k-1 is also a chain of order k, (1.6) follows.

Section 5

Proof of Theorem 4. Let $H_v(a_1, ..., a_k)$ be the entropy of the random variables $a_1(\omega), ..., a_k(\omega)$ under v. Note first that

$$\alpha_k(A) = \sup_{\substack{v \in \mathcal{M}_r \\ v_1 \in A}} \frac{H_v(a_1, \dots, a_k)}{2\int \log q_k(\omega) v(d\omega)}.$$
(5.1)

Indeed, if v lies in \mathcal{M}_r (or even \mathcal{N}_r) and $p = v_k$, then $p \in C_k$ and the ratios in (5.1) and (1.11) coincide. On the other hand, as the argument involving (4.21) shows, the supremum in (1.11) is unaltered if further restricted to p for which the corresponding Markov chain is ergodic; for such a p, if $v = v_p$, then $v \in \mathcal{M}_r$ and the ratios in (5.1) and (1.11) coincide.

By (1.2) and (2.3),

$$\left|\frac{2}{k}\int \log q_k(\omega)v(d\omega) - t(v)\right| \leq \frac{8}{k}.$$
(5.2)

And (see [3; p. 82], for example)

$$\frac{1}{k}H_{\nu}(a_1,\ldots,a_k) \ge h(\nu).$$
(5.3)

For positive ε , there is by (1.7) a v in \mathcal{M}_r with $h(v)/t(v) > \dim L_r(A) - \varepsilon$. Choose k_0 so that $h(v)/(t(v)+8/k_0) > \dim L_r(A) - \varepsilon$. By (5.2) and (5.3), $k \ge k_0$ implies

$$\alpha_k(A) \geq \frac{H_{\nu}(a_1, \ldots, a_k)}{2\int \log q_k(\omega)\nu(d\omega)} \geq \frac{h(\nu)}{t(\nu) + 8/k} > \dim L_r(A) - \varepsilon.$$

Thus $\lim \inf_k \alpha_k(A) \ge \dim L_r(A)$.

If $j \leq k$ and $v \in \mathcal{N}_r$, then by standard properties of conditional entropy,

$$H_{\nu}(a_1, \dots, a_k) = H_{\nu}(a_1, \dots, a_j) + \sum_{v=j}^{k-1} H_{\nu}(a_{v+1} | a_1, \dots, a_v)$$

$$\leq \log r^j + k H_{\nu}(a_j | a_1, \dots, a_{j-1}).$$

If v^j is the measure corresponding to that Markov chain of order j-1 whose *j*-dimensional distribution coincides with that of v (that is, if $v_i^j = v_j$), then

$$H_{\nu}(a_1, \dots, a_k) \leq j \log r + k h(\nu^j), \quad j \leq k.$$
(5.4)

By (4.15), t(v) is within $8/2^{j}$ of

$$-2\int \log\left(\frac{1}{a_1(\omega)} + \dots + \frac{1}{a_j(\omega)}\right) v(d\omega), \qquad (5.5)$$

and similarly for $t(v^j)$; but (5.5) is the same for v^j as for v, and hence $|t(v) - t(v^j)| \le 16/2^j$.

By this, (5.4), (5.2), and (4.2),

$$\frac{H_{\nu}(a_{1},...,a_{k})}{2\int \log q_{k}(\omega) \nu(d\omega)} \leq \frac{\frac{j}{k}\log r + h(\nu^{j})}{t(\nu^{j}) - \frac{8}{k} - \frac{16}{2^{j}}}$$
$$\leq \frac{\frac{j}{k}\log r}{\frac{1}{6} - \frac{8}{k} - \frac{16}{2^{j}}} + \frac{1}{1 - 6\left(\frac{8}{k} + \frac{16}{2^{j}}\right)} \cdot \frac{h(\nu^{j})}{t(\nu^{j})}$$

for $j \leq k$.

We next show that $v \in \mathcal{N}_r$ implies $v^j \in \mathcal{M}_r$. Consider cylinders

$$M_n = [\omega: a_{n+\nu}(\omega) = i_{\nu}, 0 \le \nu \le j-2]$$

and

$$M'_{n} = [\omega: a_{n+v}(\omega) = i'_{v}, 0 \leq v \leq j-2].$$

Since v is ergodic,

$$n^{-1}\sum_{k=1}^{n} v(M_1 \cap M'_k) \to v(M_1) v(M'_1) = v^j(M_1) v^j(M'_1).$$

Thus $v(M_1 \cap M'_n)$ is positive for some *n* if $v^j(M_1)$ and $v^j(M'_1)$ are both positive; but then $v^j(M_1 \cap M'_n)$ is also positive for some *n*, because v(M) > 0 implies $v^j(M) > 0$ for cylinders *M*. Therefore, if two sequences (of length j-1) of states are possible under v^j , it is also possible to pass from one to the other; hence v^j is ergodic.

It therefore follows from (1.7) that we may replace $h(v^j)/l(v^j)$ by dim $L_r(A)$ in the preceding inequality if $v \in \mathcal{N}_r$. Applying (5.1) to the left side of the inequality, letting $k \to \infty$, and then letting $j \to \infty$, we obtain $\limsup_k \alpha_k(A) \leq \dim L_r(A)$, which proves Theorem 4.

Section 6

A Conjecture. Let $\beta_k(A)$ be the supremum of the ratio in (1.11) over those p with nonnegative components satisfying (1.8) and $\tau_k(p) \in A$. In other words, drop the constraints (1.9). Then $\alpha_k(A) \leq \beta_k(A)$, and it may be conjectured that

$$\lim_{k \to \infty} (\beta_k(A) - \alpha_k(A)) = 0, \qquad (6.1)$$

so that $\alpha_k(A)$ can be replaced by $\beta_k(A)$ in (1.12).

The conjecture is true for $A = C_1$: If s satisfies

$$\sum q_k(i_1, \dots, i_k)^{-2s} = 1 \tag{6.2}$$

 $(i_1, ..., i_k \text{ ranging from } i \text{ to } r \text{ as usual})$, and if $p(i_1, ..., i_k) = q_k(i_1, ..., i_k)^{-2s}$, then p satisfies (1.8) and a computation shows that the ratio in (1.11) has the value s. On the other hand, if s satisfies (6.2) and the $p(i_1, ..., i_k)$ are nonnegative and add

to 1, then

$$\sum p(i_1, ..., i_k) \log p(i_1, ..., i_k)^{-1} q_k(i_1, ..., i_k)^{-2s}$$

$$\leq \log \sum q_k(i_1, ..., i_k)^{-2s} = 0$$

by convexity, so the ratio in (1.11) is at most s. Thus $\beta_k(C_1)$ is the root of the Eq. (6.2). Good [5] has shown that this root converges to dim $\Omega_r = \dim U_r(C_1)$, so (6.1) does hold for $A = C_1$. It would be interesting to have a simple, direct proof of (6.1), at least for this special case.

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