

GUESSING THE NEXT OUTPUT OF A STATIONARY PROCESS

BY
DONALD S. ORNSTEIN*

ABSTRACT

Suppose we start watching a stationary process at time 0. Then the conditional probability of a particular output at time -1 , given the outputs at times 0 through k , will converge. In this paper we will show that we can make a guess, depending only on the outputs from 0 through k (and not, of course, on the process) that will converge to the above limit with probability one.

Introduction

In this paper we will consider stationary processes that print out two symbols 0, or 1.** (By a stationary process we will mean a shift invariant measure on doubly infinite sequences of 0, 1 or equivalently a measure preserving transformation and a partition into two sets.) We will assume the process is ergodic.†

We will be concerned with the conditional probability that the process will print out a , 0 at time 1 given its past. This makes sense as follows: Let a_i be the printout at time i . Let $G(a_0, a_{-1}, \dots, a_{-k})$ denote the conditional probability of $a_1 = 0$ given that the printout at times 0 through $-k$ were (a_0, \dots, a_{-k}) . The Martingale convergence theorem says that with probability one $G(a_0, \dots, a_{-k})$ converges as $k \rightarrow \infty$.

If we know the infinite past we can find (with probability one) each $G(a_0, \dots, a_{-k})$ because the ergodic theorem (or law of large numbers) says that the probability of any finite string is equal (with probability one) to the frequency with which it will occur in the past.

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** (Our results hold with no added difficulties for stationary processes that print out a finite or countable number of symbols.)

† This is no real restriction because every stationary process decomposes into ergodic ones, i.e. all non ergodic processes are obtained by randomly picking an ergodic process.

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This means that if we are given the infinite past of a stationary process we can, without any further information about the process, determine the probability that the next printout will be a 0.

Thomas Cover pointed out that we can ask a more subtle question: is there a guessing scheme such that given a sequence of 0, 1 of length k there is a guess, $f(a_0, \dots, a_{-k})$ (f is some number between 0 and 1), with the property that, for any stationary process, if we apply f to the last k outputs and let k go to infinity, then with probability one (the probability is given by the process) $f(a_0, \dots, a_{-k})$ will converge to the probability that $a_1 = 0$ given the past.

The purpose of this paper is to show that such a scheme exists.

We could rephrase our problem (reversing time) as follows: Suppose we start watching a stationary process at time 0. The conditional probability of a 0 at time -1 given the printout up to time k will converge and we want to make a guess, based only on the printout up to time k , that will converge to the same thing.

The problem is especially interesting in the case of deterministic processes, or processes of 0 entropy (i.e., the conditional probability of a 0 given the past is 0 or 1). In this case it is said that we can predict the future from the past, or the past from the future. It was not commonly realized until Cover pointed it out that that statement could have many meanings. A special case of the result of this paper shows that the above is true in a strong sense, that is: if we start watching a deterministic process at time 0 (we are given no information about the process) we can, at each finite time, make a guess as to the process printed out at time -1 before we started watching and with probability one our guesses will converge to the correct answer.

We will now construct a scheme for guessing the output at -1 , given the outputs starting at 0.

We will start with some notation. Let a_0, a_1, \dots, a_k be a sequence of 0's or 1's ($a_i = 0$ or $a_i = 1$). Let b_0, \dots, b_l , $l < k$ be another such sequence. We will say that b_0, \dots, b_l occurs in a_0, \dots, a_k at a_j if $a_{j+i} = b_i$, $0 \leq i \leq l$. If $a_{j-1} = 0$ we will say that the occurrence is preceded by 0. If we are given a sequence a_0, \dots, a_1, \dots define $g(i, j)$, $i < j$ as follows: Let r denote the number of occurrences of a_0, \dots, a_i in a_0, \dots, a_j , not counting the occurrence at a_0 . Let r_0 denote the number of the above occurrences that are preceded by 0. Let $g(i, j) = r_0/r$ (the $g(i, j)$ are "reasonable" guesses for the conditional probability that $a_{-1} = 0$). We will say that a_0, \dots, a_k is (N, K, ε) acceptable if there is a subsequence $K = a_{n_0}, < a_{n_1} < \dots < a_{n_N} = a_k$ such that

(1) $|g(i, n_j) - g(s, n_t)| < \varepsilon$ if $K \leq i \leq n_{j-1}$, $K \leq s \leq n_{t-1}$ and $1 \leq j \leq N$, $1 \leq t \leq N$.

(2) If $K \leq i \leq n_{j-1}$, $1 \leq j \leq N$, then the number of occurrences of a_0, \dots, a_i in a_0, \dots, a_{n_j} is $\geq K$.

We will refer to a_0, \dots, a_k together with $K = n_0 < \dots < n_N = k$ as a *partitioned sequence* (the partition tells us which “guesses” $g(i, j)$ to consider).

We can now describe our guessing scheme. Let a_0, a_1, \dots be the outcomes of our process. Pick a sequence $\varepsilon_i \rightarrow 0$. Wait until there is an l_1 such that for some N_1 , a_0, \dots, a_{l_1} is $(N_1, N_1, \varepsilon_1)$ acceptable. Then predict $g(N_1, l_1)$. Keep making the same guess until there is an l_2 such that there is an N_2 and a_0, \dots, a_{l_2} is $(N_2, N_2, \varepsilon_2)$ etc.

LEMMA 1. *Given $\varepsilon > 0$ and M then with probability 1 there will be an L such that if $l > L$ then a_0, \dots, a_l is (N, N, ε) acceptable for some $N > M$.*

PROOF. The Martingale convergence theorem tells us that given $\varepsilon > 0$ there is an M such that if $j > M$ and $i > M$, then the conditional probability that $a_{-1} = 0$, given a_0, \dots, a_i or a_0, \dots, a_j , differ by $< \frac{1}{2}\varepsilon$. The lemma now follows from the Birkhoff ergodic theorem (law of large numbers).

Define $G(a_0, \dots, a_k)$ to be the conditional probability that a_{-1} is 0 given that the outputs of the process at times 0 through k are a_0, \dots, a_k . We will say that a_0, \dots, a_l is (N, K, α) bad if there exists $K = n_0 < n_1 < \dots < n_N = l$, and $g(i, n_j) - G(a_0, \dots, a_i) > \alpha$ for all $K \leq i \leq n_{j-1}$, $1 \leq j \leq N$ (or if all of the above numbers are $< -\alpha$).

LEMMA 2. *Either our guessing scheme converges to the right guess or there is an $\alpha > 0$ and increasing sequences l_i, N_i such that a_0, \dots, a_{l_i} is (N_i, N_i, α) bad.*

PROOF. If our guessing scheme does not work there will be an $\alpha > 0$ such that we are wrong by $> 2\alpha$ infinitely often. There is an M (Martingale convergence theorem) such that if $i > M$ and $j > M$ then

$$|G(a_0, \dots, a_i) - G(a_0, \dots, a_j)| < \frac{1}{100} \alpha.$$

If a_0, \dots, a_l is $(N, N, (1/100)\alpha)$ acceptable then all of the $g(i, j)$ that we used are within $(1/100)\alpha$ of each other. Therefore, if $N > M$, $\beta < (1/100)\alpha$ and a_0, \dots, a_l is (N, N, β) acceptable and gives a guess that is off by $> 2\alpha$ then a_0, \dots, a_l must be bad.

LEMMA 3. *Given $\alpha > 0$ there is a $\beta < 1$ such that if $1/K < (1/100)\alpha$ then the probability that there will be an l and N such that a_0, \dots, a_l is (N, K, α) bad is $< \beta^N$.*

Lemma 3 implies easily that our guessing scheme will converge with probability one. The proof of Lemma 3 will depend on the next two lemmas.

The badness condition states the existence of a sequence and a partitioning. The next lemma gives us canonical sequences and partitionings. This makes it easier to compute probabilities.

LEMMA 4. *Assume a_0, \dots, a_k is (N, K, α) bad. Let $n_i, i \leq N$ be the smallest integer such that a_0, \dots, a_{n_i} is (i, K, α) bad. Then a_0, \dots, a_{n_N} is (N, K, α) bad with respect to the partition $K = n_0 < n_1 < \dots < n_N = l$.*

PROOF. We must check that

(a) $g(i, n_j) - G(a_0, \dots, a_i) > \alpha$ and

(b) a_0, \dots, a_i occurs $> K$ times in a_0, \dots, a_{n_j} , for all pairs $n_j, 1 \leq j \leq N$ and $i \leq n_{j-1}$. Consider one n_j . Because a_0, \dots, a_{n_j} is (j, K, α) bad there exists $K = \bar{n}_0 < \bar{n}_1 < \bar{n}_j = n_j$.

(a) and (b) must hold for n_j and all i less than \bar{n}_{j-1} . We will therefore be finished if we show $\bar{n}_{j-1} \geq n_{j-1}$. However, $a_0, \dots, a_{n_{j-1}}$ is $(j-1, K, \alpha)$ bad and therefore $\bar{n}_{j-1} \geq n_{j-1}$.

LEMMA 5. *Given $\alpha > 0$ there exists a $\beta < 1$ with the following property. Fix a sequence b_0, \dots, b_n . Let $p(b_0, \dots, b_n)$ denote the conditional probability of printing a sequence $a_0, \dots, a_l, l > n$ given that $a_i = b_i, 0 \leq i \leq n$ and such that (1) $g(n, l) - G(a_0, \dots, a_n) > \alpha$ and (2) a_0, \dots, a_n occurs more than $100/\alpha$ times in a_0, \dots, a_l . Then $p(b_0, \dots, b_n) < \beta$.*

PROOF. Using the ergodic theorem we can pick an infinite sequence c_0, c_1, \dots with the property that for any finite sequence d_1, \dots, d_j , the frequency of occurrences of d_1, \dots, d_j will equal the probability that the i th output is $d_i, 1 \leq i \leq j$. We will therefore compute probabilities by computing frequencies of occurrences in c_0, c_1 .

It will be convenient to prove our lemma with l restricted to be $<$ some M ; and note that for all M we get the same β .

Look at all of the occurrences of b_0, \dots, b_n . We will call an occurrence special if it can be extended to the occurrence of a sequence a_0, \dots, a_l satisfying (1), (2), and $l < M$. Because of (1) and (2) the occurrences of b_0, \dots, b_n in the occurrence of a_0, \dots, a_l (including the first) will form a collection of consecutive occurrences in which the percentage of occurrences preceded by 0 is too high. (We need (2) because we are including the first occurrence.) Call such a collection a special collection. We now take the next special occurrence of b_0, \dots, b_n that occurs to the right of the previous special collection and form the special collection

starting with it. (This will be disjoint from our previous special collection.) If we proceed in this way, we will get a set, S , of disjoint, special collections, which include all of the special occurrences. Since the percentage of occurrences preceded by 0, in S , is greater than the percentage of occurrences preceded by 0 in the whole string, it follows that a certain percentage of occurrences will not be in S and will therefore not be special. (Note that this percentage can be taken to be independent of M .)

PROOF OF LEMMA 3. We will show that the probability of printing out an $(N + 1, K, \alpha)$ bad sequence is $< \beta$ times the probability of an (N, K, α) bad sequence. Partition the infinite sequences, that have an (N, K, α) bad initial segment, according to their canonical (N, K, α) bad sequence a_0, \dots, a_{n_N} defined in Lemma 4. Call these classes C_i . In each C_i the probability (by Lemma 5) of extending a_0, \dots, a_{n_N} to some a_0, \dots, a_l , where a_0, \dots, a_{n_N} occurs more than $100/\alpha$ times in a_0, \dots, a_l and $|G(a_0, \dots, a_{n_N}) - g(n_N, l)| > \alpha$, is less than β . But by Lemma 4 if a_0, \dots, a_{n_N} extended to an $(N + 1, K + \alpha)$ bad sequence it would have an extension of the above form.

DEPARTMENT OF MATHEMATICS
STANFORD UNIVERSITY
STANFORD, CALIFORNIA 94305 USA