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ON STOCHASTIC APPROXIMATION METHODS¹

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In [1] A. Dvoretzky proved the theorem quoted below, which implies all previous results on the convergence to a limit of stochastic approximation methods. (For a description of these results see [1].) In the present note we give a simple and, we think, perspicuous proof of this theorem which may be of help in further work. The present note is entirely self-contained and may be read without reference to [1].

Theorem. (Dvoretzky) Let α_n , β_n and $\gamma_n(n = 1, 2, \cdots)$ be non-negative real numbers satisfying

$$\lim_{n\to\infty}\alpha_n=0,$$

(2)
$$\sum_{n=1}^{\infty} \beta_n < \infty,$$

and

(3)
$$\sum_{n=1}^{\infty} \gamma_n = \infty.$$

Let θ be a real number and $T_n(n=1,2,\cdots)$ be measurable transformations satisfying

$$(4) |T_n(r_1, \dots, r_n) - \theta| \leq \max[\alpha_n, (1 + \beta_n)|r_n - \theta| - \gamma_n]$$

for all real r_1, \dots, r_n . Let X_1 and $Y_n(n = 1, 2, \dots)$ be random variables and

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$$T_n(X_1, \cdots, X_n),$$

just as is done in [1]. No ambiguity will be caused by this.

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² In the proof of the theorem we will, for the sake of brevity, write $T_n(X_n)$ for

$$(5) X_{n+1}(\omega) = T_n(X_1(\omega), \cdots, X_n(\omega)) + Y_n(\omega)$$

for $n \geq 1$.

Then the conditions $E\{X_1^2\} < \infty$,

$$(6) \qquad \qquad \sum_{n=1}^{\infty} E\{Y_n^2\} < \infty$$

and

(7)
$$E\{Y_n \mid X_1, \dots, X_n\} = 0$$

with probability 1 for all n, imply

(8)
$$\lim_{n=\infty} E\{(X_n - \theta)^2\} = 0$$

and

$$(9) P\{\lim_{n\to\infty} X_n = \theta\} = 1.$$

EXTENSION. The theorem remains valid if α_n and β_n in (4) are replaced by non-negative functions $\alpha_n(r_1, \dots, r_n)$ and $\beta_n(r_1, \dots, r_n)$ respectively, provided: The functions $\alpha_n(r_1, \dots, r_n)$ are uniformly bounded and

(10)
$$\lim_{n\to\infty}\alpha_n(r_1,\,\cdots,\,r_n)\,=\,0$$

uniformly for all sequences r_1, \dots, r_n, \dots ; the functions $\beta_n(r_1, \dots, r_n)$ are measurable and

(11)
$$\sum_{n=1}^{\infty} \beta_n(r_1, \cdots, r_n)$$

is uniformly bounded and uniformly convergent for all sequences r_1, \dots, r_n, \dots ; and for any L > 0 there exist non-negative functions $\gamma_n(r_1, \dots, r_n)$ satisfying (4), and

(12)
$$\sum_{n=1}^{\infty} \gamma_n(r_1, \dots, r_n) = \infty$$

holds uniformly for all sequences r_1 , \cdots , r_n , \cdots for which

$$\sup_{n=1,2,\cdots} |r_n| < L.$$

Proof: Without loss of generality we may take $\theta = 0$.

I. From (4) and (6) it follows readily that $EX_n^2 < \infty$ for any n.

II. Define s(n) to be the sign of $[T_n(X_n)][X_n]$ if neither factor is zero, and s(n)=1 if either factor is zero. Define $\pi(m,n)=\prod_{j=m}^n s(j),\ Y'_n=\pi(1,n)Y_n$. The series $\sum_{j=1}^{\infty} Y'_n$ converges w.p.1, by Loève ([2], p. 387, D) and (6) and (7). Let

$$Z(m, n) = \sum_{j=m}^{n} Y'_{j}$$

For any δ and ϵ both >0, there exists $M'(\delta, \epsilon)$ such that

(14)
$$P\left\{\sup_{\substack{m,n\\M'\leq m\leq n}}|Z(m,n)|>\frac{\delta}{48}\right\}<\frac{\epsilon}{2}.$$

III. Let d(m, m - 1) = 1 and, for $n \ge m$,

$$d(m, n) = \prod_{i=m}^{n} (1 + \beta_i).$$

Consider the sum

$$S(m, n) = \sum_{j=m}^{n+1} d(j, n) Y'_{j-1},$$

which is equal to

(15)
$$\sum_{j=m}^{n-1} Z((m-2), (j-1))[d(j,n) - d(j+1,n)] - Y'_{m-2} d(m,n) + Z((m-2), (n-1)) d(n,n) + Y'_{n}$$

Since $d(j, n) \ge d(j + 1, n)$ we have that the absolute value of (15) is not greater than

$$2 \left[\sup_{m-1 \le j \le n} |Z((m-2), (j-1))| \right] (d(m, n)) + |Y_n|.$$

Hence, from (11) and (14) it follows that, for δ and ϵ both >0, there exists an $M''(\delta, \epsilon) \ge M'(\delta, \epsilon)$ such that $d(m, \infty) < \frac{3}{2}$ for $m \ge M''$ and

$$(16) \qquad P\left\{\sup_{\substack{m,n\\M''\leq m\leq n}}|Z(m,n)|<\frac{\delta}{48}, \sup_{\substack{m,n\\M''\leq m\leq n}}|S(m,n)|<\frac{\delta}{8}\right\}>1-\frac{\epsilon}{2}.$$

Proof of (9) under the conditions of the extension. Let ϵ and δ be positive and arbitrary. It is sufficient to prove that

(17)
$$P\{|X_n| < \delta \text{ for all } n \text{ sufficiently large}\} > 1 - \epsilon.$$

Let $M \ge M''(\delta, \epsilon)$ be so large that, for $n \ge M$, $\alpha_n < \delta/8$. Let L be so large that $L > \delta$ and

(18)
$$\max_{1 \le j \le M} EX_j^2 < \frac{\epsilon L^2}{32M}.$$

We take this to be the L for which (12) holds. It also follows that

(19)
$$P\left\{\max_{1\leq j\leq M} |X_j| \leq \frac{L}{4}\right\} > 1 - \frac{\epsilon}{2}.$$

Suppose that the following four conditions are fulfilled:

$$|X_m| \leq \frac{\delta}{4} \quad \text{for some} \quad m \geq M;$$

$$(22) |X_{m+j}| > \frac{\delta}{4}, \quad 1 \le j \le k;$$

$$(23) |X_{m+k+1}| \leq \frac{\delta}{4}.$$

Here $1 \le k \le \infty$. In case $k = \infty$, (22) is to hold for all $j \ge 1$ and (23) is vacuous. (It will be clear by the time the proof is finished that k cannot $= \infty$.) Because $\alpha_n < \delta/8$ for $n \ge M$ and because of (20), (21), and (22) it follows that

$$(24) |T_{m+j}(X_{m+j})| > \alpha_{m+j}, 0 \le j \le k-1,$$

(25)
$$\operatorname{sign} X_{m+j+1} = \operatorname{sign} T_{m+j}(X_{m+j}), \quad 0 \le j \le k-1.$$

Applying (4) (with the γ 's zero) we obtain that X_{m+1} -lies between zero and

(26)
$$s(m)(1 + \beta_m)X_m + Y_m.$$

Repeating this argument, we obtain that, for $1 \leq j \leq k$, X_{m+j} lies between 0 and

$$s(m+j-1)s(m+j-2)\cdots s(m) d(m, m+j-1)X_m$$

$$(27) + s(m+j-1) \cdots s(m+1) d(m+1, m+j-1) Y_m + \cdots + s(m+j-1) d(m+j-1, m+j-1) Y_{m+j-2} + Y_{m+j-1}.$$

The absolute value of (27) is not greater than

$$|X_m| d(m, m+j-1) + |S(m+1, m+j-1)|.$$

Hence

$$(29) |X_{m+j}| < \delta, 1 \le j \le k.$$

To prove (17) it remains only to show that the following conditions cannot both hold:

(30) the relations in curly brackets in (16) and (19);

$$|X_n| > \frac{\delta}{4} \quad \text{for all} \quad n \ge M.$$

Applying the argument of the previous paragraph with δ replaced by L we obtain that

$$|X_n| < L \text{ for all } n \ge 1.$$

Hence (12) holds. In view of (30) and (31) it follows that

$$|T(X_n)| > \alpha_n \text{ for all } n \ge M - 1,$$

(34)
$$\operatorname{sign} T_n(X_n) = \operatorname{sign} X_{n+1} \text{ for all } n \ge M - 1.$$

We may now, and do, apply the argument which led to (28), but with the γ 's which satisfy (12). We conclude that, for all n > M, the absolute value of $|X_n|$ is not greater than

(35)
$$|X_M| d(M, n-1) + |S(M+1, n-1)| - \sum_{i=M}^{n-1} \gamma_i$$

For n sufficiently large this becomes negative, contradicting (33) and hence (31). This completes the proof of (9).

The fact that $EX_1^2 < \infty$ is used in the above proof only in order that $EX_n^2 < \infty$ for all n, and this latter fact is needed only for (8), and not for (9). For in the proof above we used the fact that $EX_n^2 < \infty$ only to obtain explicitly an L for which (19) holds. Such an L obviously exists whether or not $EX_n^2 < \infty$.

Proof of (8) under the conditions of the extension. Let $K = \max_{1 \le j < \infty} \alpha_j$. Let N be an integer to be chosen later. In view of (9) we have only to prove that $\lim_{n\to\infty} E\{(|X_n|-K)^+\}^2=0$. Let P denote probability measure and A be any set in the sample space which can be defined in terms of X_1, \dots, X_m . We use the inequality

$$H_{m+1}(A) = \int_{A} ((|X_{m+1}| - K)^{+})^{2} dP = \int_{A} ((|T_{m}(X_{m}) + Y_{m}| - K)^{+})^{2} dP$$

$$\leq \int_{A} [Y_{m}^{2} + ((|T_{m}(X_{m})| - K)^{+})^{2}] dP$$

$$\leq \int_{A} [Y_{m}^{2} + K\beta_{m}(1 + K\beta_{m}) + (1 + \beta_{m})^{2}(1 + K\beta_{m})((|X_{m}| - K)^{+})^{2}] dP$$

which is in [1] and can be deduced from (4) and (7). Let B(j) be the set $\{|X_{N+j}| \le K, |X_{N+i}| > K \text{ for } 0 \le i < j\}$, D(j) the complement of

$$B(0) + B(1) + \cdots + B(j).$$

Iterate the inequality (36) to obtain an upper bound on $H_n(A)$, n > N, beginning the iteration at $m = N, N + 1, \dots, n - 1$, respectively, and using as A the sets B(0), B(1), \dots , B(n - N - 1), respectively. In each case the last term of the integrand of the right member of (36) vanishes. Adding, we obtain that $H_n(B(0) + \dots + B(n - N))$ can be made arbitrarily small by making N sufficiently large.

It remains only to consider $H_n(D(n-N))$. For any point in D(n-N) we have, as in (27), that

$$|X_n| \leq |\pi(1, N-1) d(N, n-1) X_N + S(N+1, n-1)|$$

Hence, by Minkowski's inequality

(38)
$$\left(\int_{D(n-N)} (X_n)^2 dP \right)^{\frac{1}{2}}$$

$$\leq [d(1, \infty)] \left(\int_{D(n-N)} (X_N)^2 dP \right)^{\frac{1}{2}} + [d(1, \infty)] \left(\sum_{j=N}^{\infty} EY_j^2 \right)^{\frac{1}{2}}$$

The second term on the right of (38) can be made arbitrarily small by making N sufficiently large. The first term can be made arbitrarily small by making n sufficiently large, since $P\{D(n-N)\} \to 0$ as $n \to \infty$. This completes the proof of (8).

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ON THE DERIVATIVES OF A CHARACTERISTIC FUNCTION AT THE ORIGIN

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1. Introduction. Let F(x), $-\infty < x < \infty$, be a distribution function, and

$$\phi(t) = \int_{-\infty}^{\infty} e^{itx} dF(x)$$

its characteristic function, defined and continuous for all real t. Let k be a positive integer. If the kth moment of F(x),

$$\mu_k = \int_{-\infty}^{\infty} x^k dF(x),$$

exists and is finite (integral absolutely convergent), $\phi(t)$ has a finite kth derivative for all real t given by

$$\phi^{(k)}(t) = i^k \int_{-\infty}^{\infty} x^k e^{itx} dF(x).$$

In particular,

$$\phi^{(k)}(0) = i^k \mu_k.$$

The existence and finiteness of μ_k is a sufficient condition for the existence and finiteness of $\phi^{(k)}(0)$. It can be shown (see [1]) that when k is even, this condition is also necessary; but when k is odd this is not so. Zygmund [2] has given a necessary and sufficient condition for the existence of $\phi'(0)$ and also one for the existence of a symmetric derivative of higher odd order at t=0; but he imposes a certain condition (smoothness) on the characteristic function. In the following theorem the conditions are on the distribution function only.