

STOCHASTIC ESTIMATION OF THE MAXIMUM OF A REGRESSION FUNCTION¹

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1. Summary. Let $M(x)$ be a regression function which has a maximum at the unknown point θ . $M(x)$ is itself unknown to the statistician who, however, can take observations at any level x . This paper gives a scheme whereby, starting from an arbitrary point x_1 , one obtains successively x_2, x_3, \dots such that x_n converges to θ in probability as $n \rightarrow \infty$.

2. Introduction. Let $H(y | x)$ be a family of distribution functions which depend on a parameter x , and let

$$(2.1) \quad M(x) = \int_{-\infty}^{\infty} y dH(y | x).$$

We suppose that

$$(2.2) \quad \int_{-\infty}^{\infty} (y - M(x))^2 dH(y | x) \leq S < \infty,$$

and that $M(x)$ is strictly increasing for $x < \theta$, and $M(x)$ is strictly decreasing for $x > \theta$. Let $\{a_n\}$ and $\{c_n\}$ be infinite sequences of positive numbers such that

$$(2.3) \quad c_n \rightarrow 0,$$

$$(2.4) \quad \sum a_n = \infty,$$

$$(2.5) \quad \sum a_n c_n < \infty,$$

$$(2.6) \quad \sum a_n^2 c_n^{-2} < \infty.$$

(For example, $a_n = n^{-1}$, $c_n = n^{-1/3}$.)

We can now describe a recursive scheme as follows. Let z_1 be an arbitrary number. For all positive integral n we have

$$(2.7) \quad z_{n+1} = z_n + a_n \frac{(y_{2n} - y_{2n-1})}{c_n},$$

where y_{2n-1} and y_{2n} are independent chance variables with respective distributions $H(y | z_n - c_n)$ and $H(y | z_n + c_n)$. Under regularity conditions on $M(x)$ which we shall state below we will prove that z_n converges stochastically to θ (as $n \rightarrow \infty$).

The statistical importance of this problem is obvious and need not be discussed. The stimulus for this paper came from the interesting paper by Robbins and Monro [1] (see also Wolfowitz [2]).

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While we have no need to postulate the existence of the derivative of $M(x)$ (indeed, $M(x)$ can be discontinuous), the spirit of our regularity assumptions postulated below is as follows. (a) If $M(x)$ did have a derivative it would be zero at $x = \theta$. Hence we would have expected the derivative not to be too large in a neighborhood of $x = \theta$. (b) If, at a distance from θ , $M(x)$ were very flat, then movement towards θ would be too slow. Hence outside of a neighborhood of $x = \theta$ we would have liked the absolute value of the derivative to be bounded below by a positive number. (c) If $M(x)$ rose too steeply in places we might through mischance get a movement of z_n which would throw us far out from θ . If there were many such steep places z_n could be made to approach $+\infty$ or $-\infty$ with positive probability. We would therefore have postulated a Lipschitz condition.

From the mathematical point of view it would be aesthetic to weaken the conditions. From the practical point of view it might be objected that these conditions prevent $M(x)$ from being a function which flattens out toward the x -axis, for example, $M(x) = e^{-x^2}$, or from being a function which drops off steadily faster to $-\infty$, for example, $M(x) = -x^2$. Now in any practical situation one can always give a priori an interval $[C_1, C_2]$ such that $C_1 \leq \theta \leq C_2$. It will be sufficient if our conditions are fulfilled in this interval.

Suppose, however, that some $z_n \pm c_n$ falls outside the interval $[C_1, C_2]$ and one cannot take an observation at that level. If one then moves z_n so that the offending $z_n \pm c_n$ is at C_1 or C_2 , as the case may be, and proceeds as directed by (2.7), then our conclusion remains valid.

We postulate the following regularity conditions on $M(x)$.

CONDITION 1. There exist positive β and B such that

$$(2.8) \quad |x' - \theta| + |x'' - \theta| < \beta \text{ implies } |M(x') - M(x'')| < B|x' - x''|.$$

CONDITION 2. There exist positive ρ and R such that

$$(2.9) \quad |x' - x''| < \rho \text{ implies } |M(x') - M(x'')| < R.$$

CONDITION 3. For every $\delta > 0$ there exists a positive $\pi(\delta)$ such that

$$(2.10) \quad |z - \theta| > \delta \text{ implies } \inf_{\delta > \epsilon > 0} \frac{|M(z + \epsilon) - M(z - \epsilon)|}{\epsilon} > \pi(\delta).$$

3. Proof that z_n converges stochastically to 0. Let

$$(3.1) \quad b_n = E(z_n - \theta)^2,$$

$$(3.2) \quad U_n(z) = (z - \theta) E\{y_{2n} - y_{2n-1} | z_n = z\},$$

$$(3.3) \quad U_n^+(z) = \frac{1}{2}(U_n(z) + |U_n(z)|), \quad U_n^-(z) = \frac{1}{2}(U_n(z) - |U_n(z)|),$$

$$(3.4) \quad P_n = E(U_n^+(z_n)), \quad N_n = E(U_n^-(z_n)),$$

$$(3.5) \quad e_n = E(y_{2n} - y_{2n-1})^2.$$

From (2.7) we have

$$(3.6) \quad b_{n+1} = b_n + 2 \frac{a_n}{c_n} (P_n + N_n) + \frac{a_n^2}{c_n^2} e_n.$$

Adding the expressions obtained from (3.6) for $b_{j+1} - b_j$ for $1 \leq j \leq n$, we obtain

$$(3.7) \quad b_{n+1} = b_1 + 2 \sum_{j=1}^n \frac{a_j}{c_j} P_j + 2 \sum_{j=1}^n \frac{a_j}{c_j} N_j + \sum_{j=1}^n \frac{a_j^2}{c_j^2} e_j.$$

Noting that $U_n^+(z) \geq 0$ and that $U_n^+(z) > 0$ implies that $|z - \theta| < c_n$ because $M(x)$ is monotonic for $x < \theta$ and for $x > \theta$, it follows from (2.8) that, for all n for which $c_n < \frac{1}{2}\beta$, we have

$$(3.8) \quad 0 \leq U_n^+(z) < 2B c_n^2$$

It follows from (2.5) and (3.8) that the positive-term series

$$(3.9) \quad \sum_{n=1}^{\infty} \frac{a_n}{c_n} P_n$$

converges, say to α . From (2.9) we have

$$(3.10) \quad [M(z_n + c_n) - M(z_n - c_n)]^2 < R^2$$

for n sufficiently large. Also for large enough n ,

$$(3.11) \quad \begin{aligned} & E\{(y_{2n} - y_{2n-1})^2 | z_n\} \\ &= E\{(y_{2n} - M(z_n + c_n))^2 + (y_{2n-1} - M(z_n - c_n))^2 | z_n\} \\ &\quad + [M(z_n + c_n) - M(z_n - c_n)]^2 \leq 2S + R^2 \end{aligned}$$

by (2.2) and (3.10). Hence for large enough n

$$(3.12) \quad E[y_{2n} - y_{2n-1}]^2 \leq 2S + R^2.$$

Consequently from (2.6) we obtain that the positive-term series

$$(3.13) \quad \sum_{n=1}^{\infty} \frac{a_n^2}{c_n^2} e_n$$

converges, say to γ . Hence, since $b_{n+1} \geq 0$, it follows from (3.7) that

$$(3.14) \quad 2 \sum_{j=1}^n \frac{a_j}{c_j} N_j \geq -b_1 - 2\alpha - \gamma > -\infty,$$

so that the negative-term series

$$(3.15) \quad \sum_{n=1}^{\infty} \frac{a_n}{c_n} N_n$$

converges.

Let

$$(3.16) \quad K_n = \left| \frac{M(z_n + c_n) - M(z_n - c_n)}{c_n} \right|.$$

Then

$$(3.17) \quad E\{K_n | z_n - \theta | \} = \frac{P_n - N_n}{c_n}.$$

From the convergence of (3.9) and (3.15) and the divergence of $\sum a_n$, it follows that

$$(3.18) \quad \liminf_{n \rightarrow \infty} E\{K_n | z_n - \theta | \} = 0.$$

Let $n_1 < n_2 < \dots$ be an infinite sequence of positive integers such that

$$(3.19) \quad \lim_{j \rightarrow \infty} E\{K_{n_j} | z_{n_j} - \theta | \} = 0.$$

We assert that $(z_{n_j} - \theta)$ converges stochastically to zero as $j \rightarrow \infty$. For if not, there would exist two positive numbers δ and ϵ and a subsequence $\{t_j\}$ of $\{n_j\}$ such that, for all j ,

$$(3.20) \quad P\{|z_{t_j} - \theta| > \delta\} > \epsilon,$$

which implies that

$$(3.21) \quad E\{K_{t_j} | z_{t_j} - \theta | \} \geq \delta \epsilon \pi \left(\frac{\delta}{2} \right) > 0$$

for all j for which $c_{t_j} < \frac{1}{2}\delta$. But (3.21) contradicts (3.19) and the stochastic convergence to zero of $(z_{n_j} - \theta)$ is proved.

Let η and ϵ be arbitrary positive numbers. The proof of the theorem will be complete if we can show the existence of an integer $N(\eta, \epsilon)$ such that

$$(3.22) \quad P\{|z_n - \theta| > \eta\} \leq \epsilon \text{ for } n > N(\eta, \epsilon).$$

Let s be a positive number such that

$$(3.23) \quad \frac{s^2 + s}{\eta^2} < \frac{\epsilon}{2}.$$

Because z_{n_j} converges stochastically to θ there exists an integer N_0 such that

$$(3.24) \quad P\{|z_{N_0} - \theta| \geq s\} < \frac{\epsilon}{2}.$$

We may also choose N_0 so large that

$$(3.25) \quad c_n < \min\left(\frac{\rho}{2}, \frac{\beta}{2}\right) \text{ for all } n \geq N_0,$$

and

$$(3.26) \quad \sum_{n=N_0}^{\infty} \frac{a_n^2}{c_n^2} < \frac{s}{2R^2 + 4S},$$

and

$$(3.27) \quad \sum_{n=N_0}^{\infty} a_n c_n < \frac{s}{8B}.$$

Proceeding in a manner similar to that used to obtain (3.7), we have, for each $n > N_0$,

$$(3.28) \quad \begin{aligned} E\{(z_n - \theta)^2 | z_{N_0} = z\} &= (z - \theta)^2 + 2 \sum_{j=N_0}^{n-1} \frac{a_j}{c_j} E\{U_j | z_{N_0} = z\} \\ &+ \sum_{j=N_0}^{n-1} \frac{a_j^2}{c_j^2} E\{(y_{2j} - y_{2j-1})^2 | z_{N_0} = z\} \\ &\leq (z - \theta)^2 + 2 \sum_{j=N_0}^{\infty} \frac{a_j}{c_j} E\{U_j^+ | z_{N_0} = z\} + (R^2 + 2S) \sum_{j=N_0}^{\infty} \frac{a_j^2}{c_j^2} < (z - \theta)^2 + s. \end{aligned}$$

Using (3.23), (3.28), and Tchebycheff's inequality, we have

$$(3.29) \quad P\{|z_n - \theta| > \eta \mid |z_{N_0} - \theta| < s\} < \frac{\epsilon}{2}.$$

The inequalities (3.24) and (3.29) show that (3.22) holds for $N(\eta, \epsilon) = N_0$, and the proof is complete.

4. Further problems. The following remarks about further problems apply also to [1].

A. An obvious problem is to determine sequences $\{c_n\}$ and $\{a_n\}$ which would be optimal in some reasonable sense.

B. An important problem is to determine a stopping rule, that is, a rule by which the statistician decides when he is sufficiently close to θ .

C. This problem is a combination of B and a generalization of A, that is, to determine an optimal procedure with its stopping rule.

REFERENCES

- [1] H. ROBBINS AND S. MONRO, "A stochastic approximation method," *Annals of Math. Stat.*, Vol. 22 (1951), pp. 400-407.
- [2] J. WOLFOWITZ, "On the stochastic approximation method of Robbins and Monro," *Annals of Math. Stat.*, Vol. 23 (1952), pp. 457-461.