

ON THE STOCHASTIC APPROXIMATION METHOD OF  
ROBBINS AND MONRO<sup>1</sup>

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**1. Summary.** In their interesting and pioneering paper Robbins and Monro [1] give a method for "solving stochastically" the equation in  $x$ :  $M(x) = \alpha$ , where  $M(x)$  is the (unknown) expected value at level  $x$  of the response to a certain experiment. They raise the question whether their results, which are contained in their Theorems 1 and 2, are valid under a condition (their condition (4'), our condition (1) below) which is statistically plausible and is weaker than the condition which they require to prove their results. In the present paper this question is answered in the affirmative. They also ask whether their conditions (33), (34), and (35) (our conditions (25), (26) and (27) below) can be replaced by their condition (5'') (our condition (28) below). A counterexample shows that this is impossible. However, it is possible to weaken conditions (25), (26) and (27) by replacing them by condition (3) (abc) below. Thus our results generalize those of [1]. The statistical significance of these results is described in [1].

**2. Statement of the problem.** Let  $H(y | x)$  be a family of distribution functions which correspond to real values of the parameter  $x$ . Write

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

We postulate that

$$(1) \quad |M(x)| \leq C < \infty, \quad \int_{-\infty}^{\infty} (y - M(x))^2 dH(y | x) \leq \sigma^2 < \infty,$$

and that either

$$(2) \quad \begin{array}{ll} M(x) \leq \alpha - \delta, & \text{for } x < \theta, \\ M(x) \geq \alpha + \delta, & \text{for } x > \theta, \end{array}$$

for some  $\delta > 0$ , or else

$$(3a) \quad \begin{array}{l} M(x) < \alpha \text{ for } x < \theta, \\ M(\theta) = \alpha, \\ M(x) > \alpha \text{ for } x > \theta, \end{array}$$

and, for some positive  $\delta$ ,

$$(3b) \quad M(x) \text{ is strictly increasing if } |x - \theta| < \delta$$

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and

$$(3c) \quad \inf_{|x-\theta| \geq \delta} |M(x) - \alpha| > 0.$$

Let  $\{a_n\}$  be a sequence of positive numbers such that

$$(4) \quad \sum_{n=1}^{\infty} a_n = \infty,$$

$$(5) \quad \sum_{n=1}^{\infty} a_n^2 < \infty.$$

Let  $x_1$  be an arbitrary number. The Robbins-Monro convergence scheme is defined recursively for all  $n$  by

$$x_{n+1} = x_n + a_n(\alpha - y_n),$$

where  $y_n$  is a chance variable with distribution function  $H(y | x_n)$ . We shall prove the following

**THEOREM.** *If (1), (4), (5), and either (2) or (3) (abc) hold, then  $x_n$  converges stochastically to  $\theta$ .*

**3. Proof of the Theorem.** Let

$$\begin{aligned} b_n &= E(x_n - \theta)^2, \\ d_n &= E[(x_n - \theta)(M(x_n) - \alpha)], \\ e_n &= E \left[ \int_{-\infty}^{\infty} (y - \alpha)^2 dH(y | x_n) \right]. \end{aligned}$$

Then, from (1),

$$(6) \quad 0 \leq e_n \leq \sigma^2 + (C + |\alpha|)^2 = h \quad (\text{say}).$$

An examination of the proof of [1] shows that, since (6) is valid, the following results of [1] hold under our assumptions:

$$(7) \quad b_n \rightarrow \text{a limit, say } b, \text{ as } n \rightarrow \infty;$$

$$(8) \quad d_n \geq 0;$$

$$(9) \quad \sum_{n=1}^{\infty} a_n d_n < \infty;$$

$$(10) \quad b_{n+1} = b_1 + \sum_1^n a_j^2 e_j - 2 \sum_1^n a_j d_j.$$

From (4) and (9) we obtain

$$(11) \quad \liminf d_n = 0.$$

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

$$(12) \quad \lim d_{n_i} = 0.$$

We assert that  $x_{n_j}$  converges stochastically to  $\theta$ . If this were not so, there would exist an infinite subsequence  $t_1 < t_2 < \dots$  of the sequence of  $n_j$ , and positive numbers  $\epsilon$  and  $\eta$  such that, for all  $j$ ,

$$(13) \quad P\{|\hat{x}_{t_j} - \theta| > \eta\} > \epsilon.$$

But then for all  $t_j$  we would have

$$(14) \quad \begin{aligned} d_{t_j} &= E[(x_{t_j} - \theta)(M(x_{t_j}) - \alpha)] \\ &= E[(x_{t_j} - \theta) | M(x_{t_j}) - \alpha] \\ &\geq \epsilon\eta \inf_{|x-\theta| \geq \eta} |M(x) - \alpha|. \end{aligned}$$

From either (2) or (3) (abc) it follows that the last member of (14) is positive. This contradicts (12) and proves the assertion.

Let  $\epsilon$  and  $\eta$  be arbitrary positive numbers. Our theorem is proved if we can demonstrate the existence of an integer  $N(\eta, \epsilon)$  such that, if  $n > N(\eta, \epsilon)$ ,

$$(15) \quad P\{|x_n - \theta| > \eta\} \leq \epsilon.$$

Let  $s$  be a positive number such that

$$(16) \quad \frac{s^2 + s}{\eta^2} < \frac{\epsilon}{2}.$$

Since  $x_{n_j}$  converges stochastically to  $\theta$  there exists an integer  $N_0$  such that

$$(17) \quad P\{|x_{N_0} - \theta| \geq s\} < \frac{\epsilon}{2},$$

and

$$(18) \quad \sum_{n=N_0}^{\infty} a_n^2 < \frac{s}{2h}.$$

Define, for  $n > N_0$ ,

$$b'_n(z) = E\{(x_n - \theta)^2 | x_{N_0} = z\}.$$

From (6) and (10) we obtain

$$(19) \quad b'_n(z) \leq (z - \theta)^2 + h \sum_{j=N_0}^{\infty} a_j^2 \leq (z - \theta)^2 + \frac{s}{2}.$$

Consequently, when  $n > N_0$ ,

$$(20) \quad P\left\{ |x_n - \theta| \geq \eta \mid |x_{N_0} - \theta| < s \right\} < \frac{s^2 + s}{\eta^2} < \frac{\epsilon}{2}$$

by (19) and (16). From (17) and (20) we conclude that (15) holds with  $N(\eta, \epsilon) = N_0$ . Thus our proof is complete.

**4. Two counterexamples.**

4.1. We show that (1) ((4') in [1]) cannot be completely removed. Suppose that  $\theta = \alpha = 0$ , and define  $H(y | x)$  for  $x < 1$  in any way whatever provided only that

$$\begin{aligned} M(x) &= x, & -1 \leq x < 1, \\ M(x) &= -1, & x < -1. \end{aligned}$$

For all  $x$  such that

$$(21) \quad \sum_{n=1}^k \frac{1}{n} \leq x < \sum_{n=1}^{k+1} \frac{1}{n}$$

define  $H(y | x)$  as follows:

$$(22) \quad H(y | x) = 0, \quad y \leq -1,$$

$$(23) \quad H(y | x) = 1 - \frac{1}{k^2}, \quad -1 < y \leq 2k^2 - 1,$$

$$(24) \quad H(y | x) = 1, \quad y > 2k^2 - 1.$$

We have  $M(x) \equiv 1$  for  $x \geq 1$ . Thus  $M(x)$  satisfies

$$(25) \quad M(x) \text{ is nondecreasing,}$$

$$(26) \quad M(\theta) = \alpha,$$

$$(27) \quad \left. \frac{dM(x)}{dx} \right|_{x=\theta} > 0.$$

(These are the conditions (33), (34), and (35) of [1].)

Now let  $a_n = 1/n$  and  $x_1 > 2$ . We have

$$P \{x_{n+1} > x_n \mid y_1 = y_2 = \dots = y_n = -1\} \geq 1 - \frac{1}{n^2}$$

for all  $n > 1$ , and

$$P \{x_2 > x_1\} \geq \frac{3}{4}.$$

Hence there is a positive probability that  $x_n \rightarrow \infty$ .

4.2. We show that the condition

$$(28) \quad \begin{aligned} M(x) &< \alpha, & x < \theta, \\ M(x) &> \alpha, & x > \theta, \end{aligned}$$

(which is condition (5'') of [1]), and the condition (1) (or even the stronger condition (4) of [1], i.e., that there exist a positive constant  $C'$  such that

$$(29) \quad \int_{-C'}^{C'} dH(y | x) = 1$$

identically in  $x$ ) are not sufficient for the theorem to hold. Let  $a_n = 1/(4n)$ ,  $\theta = \alpha = 0$ . For all  $x$  we define

$$(30) \quad H(y | x) = 0, \quad y \leq M(x),$$

$$(31) \quad H(y | x) = 1, \quad y > M(x).$$

We define  $M(0) = 0$ ,  $M(1) = 1$ . Let  $x_1 = 1$ . Then  $x_2 = 3/4$ . If  $x_n = y_n$  for  $n \geq 2$  we define  $M(y_n) = (4n)^{-1}$ . Since  $\sum_{n=2}^{\infty} 1/(16n^2) < 1/2$ , we have

$$P\{x_n > 1/4\} = 1$$

for all  $n$ . We can define  $M(x)$  at points  $x$  not included in our construction above in any manner compatible with (28).

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#### REFERENCE

- [1] HERBERT ROBBINS AND SUTTON MONRO, "A stochastic approximation method," *Annals of Math. Stat.*, Vol. 22 (1951), pp. 400-407.