

ESTIMATION OF A PARAMETER OF A KNOWN LAW  
OF A PROBABILITY BY STOCHASTIC APPROXIMATIONS

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**Abstract:** We consider a stochastic approximation process in a non-empty closed convex set  $K$  of  $\mathbb{R}^k$ :  $X_{n+1} = \Pi(X_n - A_n(X_1, X_2, \dots, X_n)\Psi_n(Y_n; X_n))$ , with for each  $n$ ,  $E[\Psi_n(Y_n; \theta_n)] = 0$ , and  $\Pi$  is the projection operator on  $K$ .

We denote  $T_n$  the sub- $\sigma$ -algebra generated by the events before time  $n$ .

We prove two theorems of almost sure convergence for the process  $(X_n)$  and we give two applications for estimation of a parameter of a known law of probability.

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**Key Words:** stochastic approximation, parameter estimation, Likelihood maximum

## 1. Introduction

We define a stochastic approximation process  $(X_n)$  in a non-empty closed convex subset  $K$  of  $\mathbb{R}^k$ ; we consider:

- observable random variables  $Y_n$  in  $\mathbb{R}^p$  mutually independent;  $\mathbb{R}^k$ ;
- measurable applications  $\Psi_n$  from  $\mathbb{R}^p \times \mathbb{R}^k$  into  $\mathbb{R}^l$ ;

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- for  $n \geq 1$ , a  $(k, l)$  random matrix  $A_n$ ;
- the projection operator  $\Pi$  on  $K$ ;
- the process  $(X_n)$  in  $K$  defined recursively by

$$X_{n+1} = \Pi(X_n - A_n \Psi_n(Y_n; X_n)),$$

—  $\mathbb{R}^k$  is named parameter space,  $\mathbb{R}^p$  is named space observations, and  $\mathbb{R}^l$  is named space functions of estimate.

All random variables are defined on a probability space  $(\Omega, \mathcal{A}, P)$ . Denote  $T_n$  the sub- $\sigma$ -algebra of  $\mathcal{A}$  generated by the events before time  $n$ ;  $X_1, \dots, X_n, A_1, \dots, A_n, Y_1, \dots, Y_{n-1}$  are measurable with respect to  $T_n$ .

We give in Section 2 two almost sure convergence theorems of  $\|X_n - \theta_n\|$  to 0. An application of each theorem is given in Section 3, concerning the estimation of a parameter of a known law of probability and the estimation of a parameter of gap under convex constraints.

In the following,  $\langle \cdot, \cdot \rangle$  and  $\|\cdot\|$  are respectively the usual inner product and norm in  $\mathbb{R}^k$ ;  $A'$  denotes the transposed matrix of  $A$ ,  $\lambda_{\min}(B)$  the smallest eigenvalue of  $B$ ; the abbreviation *a.s.* means almost surely.

## 2. Lemmas

Let  $(X_n)$  be a stochastic process in a subset  $K$  of  $\mathbb{R}^k$ . Suppose:

- (H1a)  $\forall n \geq 1, \|A_n \Psi_n(Y_n; X_n)\|^2$  is integrable a.s.
- (H1b)  $\sum_1^\infty E[\|A_n \Psi_n(Y_n; X_n)\|^2 / T_n] < \infty$  a.s.
- (H2)  $\sum_1^\infty \langle X_n - \theta_n, E[A_n \Psi_n(Y_n; X_n)] \rangle^- < \infty$  a.s.
- (H3)  $\sum_1^\infty \|\theta_n - \theta_{n+1}\| < \infty$  a.s.
- (H4)  $E[\|X_1\|^2] < \infty$ .

**Lemma 1.** *Assume H1a, b and H2, 3 and 4 hold; then:*

- (1) *There exist a random positive variable  $T$  such that  $\|X_n - \theta_n\| \xrightarrow{a.s.} T$ .*
- (2)  $\sum_1^\infty \langle X_n - \theta_n, E[A_n \Psi_n(Y_n; X_n)] \rangle^+ < \infty$  a.s.

*Proof.* We have:

$$X_{n+1} - \theta_{n+1} = \Pi(X_n - A_n \Psi_n) - \Pi(\theta_{n+1}),$$

$$\begin{aligned} \|X_{n+1} - \theta_{n+1}\| &\leq \|X_n - A_n \Psi_n - \theta_{n+1}\| \leq \|X_n - \theta_n - A_n \Psi_n + \theta_n - \theta_{n+1}\|, \\ \|X_{n+1} - \theta_{n+1}\|^2 &\leq \|X_n - \theta_n\|^2 - 2 \langle X_n - \theta_n, A_n \Psi_n + \theta_{n+1} - \theta_n \rangle \\ &> + \|A_n \Psi_n + \theta_{n+1} - \theta_n\|^2, \\ \|X_{n+1} - \theta_{n+1}\|^2 &\leq \|X_n - \theta_n\|^2 - 2 \langle X_n - \theta_n, A_n \Psi_n \rangle - 2 \langle X_n - \theta_n, \theta_{n+1} - \theta_n \rangle \\ &> + 4 \|A_n \Psi_n\|^2 + 4 \|\theta_{n+1} - \theta_n\|^2, \\ E[\|X_{n+1} - \theta_{n+1}\|^2 / T_n] &\leq \|X_n - \theta_n\|^2 - 2 \langle X_n - \theta_n, E[A_n \Psi_n / T_n]^+ \rangle \\ &> + 2 \langle X_n - \theta_n, E[A_n \Psi_n / T_n]^- \rangle - 2 \langle X_n - \theta_n, \theta_{n+1} - \theta_n \rangle \\ &> + 4E[\|A_n \Psi_n / T_n\|^2 / T_n] + 4\|\theta_{n+1} - \theta_n\|^2. \end{aligned}$$

As  $|\langle a, b \rangle| \leq \|a\|^2 \|b\| + \|b\|$ , we have:

$$\begin{aligned} E[\|X_{n+1} - \theta_{n+1}\|^2 / T_n] &\leq \|X_n - \theta_n\|^2 [1 + 2\|\theta_{n+1} - \theta_n\|] - 2 \langle X_n - \theta_n, E[A_n \Psi_n / T_n]^+ \rangle \\ &> + 2 \langle X_n - \theta_n, E[A_n \Psi_n / T_n]^- \rangle + 2\|\theta_{n+1} - \theta_n\| \\ &\quad + 4E[\|A_n \Psi_n / T_n\|^2 / T_n] + 4\|\theta_{n+1} - \theta_n\|^2. \end{aligned}$$

We use in the proof the Robbins-Siegmund Lemma, see [4]:

**Lemma 2.** *Let  $(\Omega, \mathcal{A}, P)$  be a probability space and  $(T_n)$  an increasing sequence of sub- $\sigma$ -algebras of  $\mathcal{A}$ . For  $n \geq 1$ , let  $z_n, \beta_n, \xi_n$  and  $\zeta_n$  be non-negative  $T_n$ -measurable random variables such that  $E[z_{n+1} | T_n] \leq z_n(1 + \beta_n) + \xi_n - \zeta_n$ . Suppose  $\sum_1^\infty \beta_n < \infty, \sum_1^\infty \xi_n < \infty$  a.s. Then  $\lim_{n \rightarrow \infty} z_n$  exists and is finite and  $\sum_1^\infty \zeta_n < \infty$  a.s.*

Under hypothesis H1a, b and H2, 3 and 4, we have:

- (1) There exist a random positive variable  $T$  such that  $\|X_n - \theta_n\| \xrightarrow{a.s.} T$ .
- (2)  $\sum_1^\infty \langle X_n - \theta_n, E[A_n \Psi_n(Y_n; X_n)] \rangle^+ < \infty$  a.s.

□

Suppose now:

(H5a)  $\forall n \geq 1, \exists S_n : K^n \rightarrow \mathbb{R}^k, S_n$  measurable:

$$E[A_n \Psi_n(Y_n; X_n) / T_n] = S_n(X_1, X_2, \dots, X_n),$$

a.s.

(H5b)  $\forall 0 < \epsilon < 1, \forall x_1, x_2, \dots, x_{n-1}, \dots \in K$

$$\sum_1^\infty \inf_{\{x: x \in K, \epsilon \|x - \theta_n\| < \frac{1}{\epsilon}\}} \langle x - \theta_n, S_n(x_1, x_2, \dots, x_{n-1}, x) \rangle^+ = +\infty.$$

(H6)  $\|A_n \Psi_n(Y_n; X_n) - E[A_n \Psi_n(Y_n; X_n)/T_n]\| \xrightarrow{a.s.} 0$

**Lemma 3.** Assume H1a, b, H4, H6, then:  $\|X_{n+1} - X_n\| \xrightarrow{a.s.} T$

*Proof.*  $\omega \in \Omega$  is fixed throughout the proof, belonging to the intersection of the following defined a.s. convergence sets:

$$C = \{\omega \in \Omega : \|X_n(\omega) - \theta_n\| \rightarrow T(\omega)\}$$

$$C_1 = \{\omega \in \Omega : \sum_1^\infty \langle X_n(\omega) - \theta_n, S_n(X_1(\omega), X_2(\omega), \dots, X_n(\omega)) \rangle^+ < \infty\}.$$

Suppose  $T(\omega) \neq 0$ . Below  $\omega$  is omitted.

Then: By H1a, there exist  $0 < \epsilon_1 < 1$  and an integer  $N(\omega)$  such that for  $n \geq N(\omega), \epsilon_1 < \|X_n(\omega) - \theta_n\| < \frac{1}{\epsilon_1}$ .

Hence  $\sum_{N(\omega)}^\infty \langle X_n(\omega) - \theta_n, S_n(X_1(\omega), X_2(\omega), \dots, X_{n-1}(\omega), X_n(\omega)) \rangle^+ = +\infty$ , a contradiction with  $\omega \in C_1$ , then  $T(\omega) = 0$ . □

### 3. Theorems of Almost Sure Convergence

**Theorem 4.** Assume H1a, b and H2, 3 and 4, H5a, b hold; then:  $\|X_n - \theta_n\| \rightarrow 0$  a.s.

*Proof.* It is a consequence of the lemmas 1 and 3 □

Suppose now:

(H7a)  $\forall n \geq 1, \exists S_n : K \rightarrow \mathbb{R}^k,$   
 $S_n$  measurable;

$$E[A_n \Psi_n(Y_n; X_n)/T_n] = S_n(X_n) \text{ a.s.}$$

(H7b)  $\forall 0 < \epsilon < 1, S(\epsilon) = \sup_n \sup_{\{x: x \in K, \epsilon < \|x - \theta\| < \frac{1}{\epsilon}\}} \|S_n(x)\| < \infty;$

(H7c) For all  $\epsilon > 0$ , there exists  $\eta > 0$  such that  $((x_1, x_2 \in K : \|x_1 - x_2\| < \eta) \Rightarrow (\sup_n \|S_n(x_1) - S_n(x_2)\| < \epsilon))$ .

(H7d) There exist a positive integer  $q$ , a sequence of integers  $(n_l)$ , for all  $0 < \epsilon < 1$  an integer  $L(\epsilon)$  such that  $n_{l+1} \geq n_l + q$  and

$$b(\epsilon) = \inf_{l > L(\epsilon)} \inf_{\{x \in K, \epsilon < \|x - \theta\| < \frac{1}{\epsilon}\}} \sum_{j \in I_l} \langle x - \theta_j, S_j(x) \rangle^+,$$

with  $I_l = \{n_l, n_l + 1, \dots, n_{l+1} - 1\}$ ;  $\sum_1^\infty b_l(\epsilon) = +\infty$ .

**Theorem 5.** Assume H1a, b, H2, 3, 4, 6, H7a, b, c, d hold, then:  $\|X_n - \theta\| \rightarrow 0$  a.s.

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