

The Continuous Dynamic Robbins-Monro Procedure

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The proving methods developed in the book by Nevel'son and Has'minskij [4] are utilized to prove the asymptotic normality of the multidimensional continuous dynamic Robbins-Monro procedure, under assumptions similar to those usually made in the theory of stochastic approximation.

1. Suppose the zero point of a regression function is a time-varying parameter, its evolution law being known to certain extent; the Robbins-Monro stochastic approximation procedure can be then adapted to track this moving point. We shall consider the continuous-time case (investigated already by Cypkin [1] from the viewpoint of adaptive systems theory). Exploiting the proving methods developed (for the non-dynamic case) in Nevel'son and Has'minskij [3], [4], we obtain results concerning mean-square convergence, the rate of a.s. convergence, and asymptotic normality of the procedure.

2. We shall make the following assumptions:

(i) $R^0(t, x)$, $\sigma_r^0(t, x)$, $1 \leq r \leq k$, are continuous mappings of $[t_0, +\infty) \times E_t$ into E_r ; $t_0 > 0$.

(ii) For every bounded region $D \subset [t_0, +\infty) \times E_t$, there is a $K_D > 0$ such that

$$|R^0(t, x) - R^0(t, y)| + \sum_{r=1}^k |\sigma_r^0(t, x) - \sigma_r^0(t, y)| \leq K_D |x - y|$$

everywhere in D .

(iii) $x = 0$ is the unique zero point of $R^0(t, x)$ for all $t \geq t_0$.

(iv) There is a positive definite matrix C and a $\lambda > 0$ such that $(CR^0(t, x), x) \leq -\lambda(Cx, x)$, for all $x \in E_t$, $t \geq t_0$.

(v) $\sum_{r=1}^k |\sigma_r^0(t, x)| \leq K(1 + |x|)$, for all $x \in E_t$, $t \geq t_0$, and some $K > 0$.

(vi) $\xi_r(t)$, $1 \leq r \leq k$, are independent (standard) Wiener processes, consistent with a non-decreasing family $\{\mathcal{F}_t, t \geq t_0\}$ of σ -fields of events.

(vii) $Q(t)$ and $q(t)$, $\theta(t)$ are matrix-valued and vector-valued functions, respectively; Q, q continuous, θ differentiable, satisfying

$$d\theta(t)/dt = Q(t)\theta(t) + q(t), \quad t \geq t_0;$$

Q is known, $R^0, \sigma_r^0, q, \theta$ are unknown in general.

(viii) $R(t, x) = R^0(t, x - \theta(t))$, $\sigma_r(t, x) = \sigma_r^0(t, x - \theta(t))$, $1 \leq r \leq k$.

(ix) $a(t)$ is a (given) positive function, $t \geq t_0$.

(x) $X^x(t)$ is the regular solution of the stochastic differential equation

$$dX(t) = Q(t)X(t)dt + a(t)(R(t, X(t))dt + \sum_{r=1}^k \sigma_r(t, X(t))d\xi_r(t)), \quad t \geq t_0,$$

with the initial condition $X(t_0) = x$, $x \in E_{t_0}$.

This is the dynamic Robbins-Monro procedure for tracking $\theta(t)$, corresponding to a situation, when at time t , the values of $R(t, x)$ are observable with experimental errors $\sum_{r=1}^k \sigma_r(t, x) \xi_r(t)$; the term $Q(t)X(t)dt$ is a correction for trend in $\theta(t)$.

Theorem 1. Under the assumptions (i)–(x) and

$$\int_{t_0}^{\infty} a(t) dt = +\infty, \quad |Q(t)| = o(a(t)), \quad |q(t)| = o(a(t)) \quad \text{for } t \rightarrow \infty,$$

we have

$$X^x(t) - \theta(t) \rightarrow 0 \quad \text{for } t \rightarrow \infty,$$

in the mean square.

Further assume:

(xi) $a(t) = a/t^\alpha$, $a > 0$, $1/2 < \alpha < 1$.

(xii) $|Q(t)| = o(1/t^\alpha)$, $|q(t)| = O(1/t^{3\alpha/2})$, $t \rightarrow \infty$.

(xiii) $R^0(t, x) = Bx + \delta(t, x)$, $|\delta(t, x)| = o(|x|)$ for $x \rightarrow 0$, uniformly in $t \in [t_0, +\infty)$; B is a matrix such that all its eigenvalues have negative real parts.

(xiv) $\lim_{t \rightarrow \infty, x \rightarrow 0} \sigma_r^0(t, x) = s_r$ exists.

(xv) $\lim_{t \rightarrow \infty} t^{3\alpha/2} q(t) = q_\infty$ exists (with $q_\infty = 0$ if $|q(t)| = o(t^{-3\alpha/2})$).

Theorem 2. Under the assumptions (i)–(xii), we have for any γ , $0 \leq \gamma < \alpha - 1/2$,

$$t^\gamma (X^x(t) - \theta(t)) \rightarrow 0 \quad \text{a.s. for } t \rightarrow \infty.$$

Theorem 3. Under the assumptions (i)–(xv), the asymptotic distribution of $t^{\alpha/2}(X^x(t) - \theta(t))$ for $t \rightarrow \infty$ is normal with mean value $a^{-1}B^{-1}q_\infty$ and covariance matrix $a \int_0^\infty e^{Bt} S e^{B^T t} dt$, with $S = \sum_{r=1}^k s_r s_r^T$.

Remark. The conditions (vii), (xii), (xv) are satisfied especially if $\theta(t) = bt^2 + c$ (Q a known matrix of constants, b, c unknown vectors) and $\alpha = 2/3$. The differential equation (vii) then becomes $d\theta/dt = Qt^{-1}\theta(t) - Qct^{-1}$, i.e., $Q(t) = Qt^{-1}$, $q(t) = -Qct^{-1}$, $\dot{q}_\infty = -Qc$. If $Q = I$, we have the linear trend in θ : $\theta(t) = bt + c$.

3. Proof of Theorem 1. Subtract $d\theta(t)$ from both sides of the equation (x); using (vii) and denoting $Z(t) = X(t) - \theta(t)$, $z = x - \theta(t_0)$, we get

$$(1) \quad dZ(t) = Q(t)Z(t)dt - q(t)dt + a(t)(R^0(t, Z(t))dt + \sum_{r=1}^k \sigma_r^0(t, Z(t))d\xi_r(t)), \quad t \geq t_0$$

$$Z(t_0) = z.$$

Let L be the differential operator corresponding to (1):

$$(2) \quad L = \partial/\partial t + (Q(t)z - q(t) + a(t)R^0(t, z), \partial/\partial z) + (1/2)a^2(t) \sum_{r=1}^k (\sigma_r^0(t, z), \partial/\partial z)^2.$$

Putting $V(z) = (Cz, z)$, C that of (iv), we have

$$(3) \quad LV(z) = 2a(t)(CR^0(t, z), z) + 2(CQ(t)z, z) - 2(Cq(t), z) + a^2(t) \sum_{r=1}^k (C\sigma_r^0, \sigma_r^0).$$

The first term on the right is less than $-2\lambda a(t)V(z)$, according to (iv); all the other terms are bounded by $b(t)(1 + V(z))$, with $b(t) = o(a(t))$, which follows from $|Q| = o(a(t))$, $|q| = o(a(t))$, from (v) and from the inequality $|z| \leq 1 + |z|^2$. Hence,

$$LV(z) \leq -\lambda a(t)V(z) + b(t), \quad t \geq t_1,$$

$$\int_{t_0}^\infty a(t)dt = +\infty, \quad b(t) = o(a(t)), \quad V(z) \geq K(z, z).$$

(Here, as well as in the sequel, K with or without subscript will denote positive constants, possibly of different values in different formulas.)

According to Lemma 1.2 in Nevel'son, Has'minskij [3], the assertion of Theorem 1 follows.

Proof of Theorem 2. The first term on the right hand side of (3) is now less than $-2\lambda a t^{-\alpha} V(z)$, the second one is bounded by $a(t) t^{-\alpha} V(z)$ with $a(t) \searrow 0$, and the fourth one by $K t^{-2\alpha}(1 + V(z))$; we have used (xi) and (xii). Using the inequality

$$(4) \quad |z| \leq \delta^{-1} t^{-\alpha/2} + \delta t^{\alpha/2} |z|^2, \quad \delta > 0,$$

and (xii), we obtain a bound for the third term:

$$2|(Cq(t), z)| \leq K_1 t^{-2\alpha} + \delta K_2 t^{-\alpha} V(z),$$

K_2 independent of δ ; choosing δ sufficiently small, we get

$$(5) \quad L V(z) \leq -\lambda a t^{-\alpha} V(z) + K_3 t^{-2\alpha}, \quad t \geq t_1.$$

Now put $V_1(t, z) = t^{2\gamma} V(z) + t^{-\varepsilon}$ where

$$(6) \quad 0 < \gamma < \alpha - 1/2, \quad 0 < \varepsilon < 2(\alpha - \frac{1}{2} - \gamma).$$

Obviously,

$$L V_1(t, z) = t^{2\gamma} L V(z) + 2\gamma t^{2\gamma-1} V(z) - \varepsilon t^{-\varepsilon-1};$$

inserting (5) for $L V(z)$, we have

$$L V_1(t, z) \leq -\lambda a t^{2\gamma-\alpha} V(z) + K_3 t^{2\gamma-2\alpha} + 2\gamma t^{2\gamma-1} V(z) - \varepsilon t^{-\varepsilon-1}.$$

The sum of terms containing $V(z)$ is negative for $t \geq t_1$, since (xi) implies $2\gamma - \alpha > > 2\gamma - 1$, and so is the sum of the remaining two terms, since (6) implies $-\varepsilon - 1 > > 2\gamma - 2\alpha$. Hence, $L V_1(t, z) < 0, t \geq t_2$.

According to Nevel'son Has'minskij [4], Corollary 3.8.1., $\{V_1(t, Z(t)), \mathcal{F}_t\}$ is a nonnegative supermartingale, which implies the a.s. existence of finite $\lim_{t \rightarrow \infty} V_1(t, Z(t))$, i.e., of finite $\lim_{t \rightarrow \infty} t^{2\gamma} V(Z(t))$. Hence, $t^{2\gamma} |Z(t)|^2 \rightarrow 0$ a.s., which entails the assertion of Theorem 2.

Proof of Theorem 3. Owing to the uniformity condition in (xiii), there are $\varepsilon > 0$ and $K > 0$ such that $|R^0(t, z)| \leq K$ for all $|z| \leq \varepsilon$ and $t \geq t_0$. Let ε be chosen in such a way that also $|\sigma_r^0(t, z)| \leq K_1$ for all $|z| \leq \varepsilon$ and $t \geq t_0$; this can be done owing to (i) and (xiv). With this ε , define (for $t \geq t_0$)

$$(7) \quad \begin{aligned} \hat{R}(t, z) &= \begin{cases} R^0(t, z), & |z| \leq \varepsilon, \\ R^0(t, \varepsilon z/|z|) |z|/\varepsilon, & |z| > \varepsilon; \end{cases} \\ \hat{\sigma}_r(t, z) &= \begin{cases} \sigma_r^0(t, z) & |z| \leq \varepsilon, \\ \sigma_r^0(t, \varepsilon z/|z|), & |z| > \varepsilon, \end{cases} \quad 1 \leq r \leq k; \end{aligned}$$

$$\delta(t, z) = \hat{R}(t, z) - Bz.$$

Together with (1), consider the auxiliary equation

$$(8) \quad d\hat{Z}(t) = Q(t) \hat{Z}(t) dt - q(t) dt + at^{-\alpha}(\hat{R}(t, \hat{Z}(t))) dt + \sum_{r=1}^k \hat{\sigma}_r(t, \hat{Z}(t)) d\xi_r(t), \\ t \geq s (\geq t_0),$$

with the initial condition $\hat{Z}(s) = \zeta$, ζ being a \mathcal{F}_s -measurable random variable, $E|\zeta|^2 < +\infty$. The corresponding differential operator is

$$L = \partial/\partial t + (Q(t)z - q(t) + at^{-\alpha}\hat{R}(t, z), \partial/\partial z) + (1/2)a^2t^{-2\alpha} \sum_{r=1}^k (\hat{\sigma}_r(t, z), \partial/\partial z)^2.$$

Put $V(z) = (Cz, z)$; we have as in (5)

$$LV(z) \leq -\lambda a V(z) + Kt^{-2\alpha}, \quad t \geq t_1;$$

hence (see Nevel'son, Has'minskij [4], formula 3.5.5, which is valid here, owing to the definition of \hat{R} , $\hat{\sigma}_r$)

$$\frac{d}{dt} E V(\hat{Z}^s(t)) = EL V(\hat{Z}^s(t)) \leq -\lambda at^{-\alpha} E V(\hat{Z}^s(t)) + Kt^{-2\alpha}, \quad t \geq s.$$

From this differential inequality, we get (see Lemmas 1, 2, 3 in Dupač [2])

$$(9) \quad E|\hat{Z}^s(t)|^2 \leq K_1 t^{-\alpha}.$$

Denoting $\hat{Y}(t) = t^{\alpha/2} \hat{Z}^s(t)$, we obtain from (8) the equation

$$d\hat{Y} = (\frac{1}{2}\alpha It^{-1} + aBt^{-\alpha})\hat{Y} dt + Q(t) \hat{Z} t^{\alpha/2} dt - q(t) t^{\alpha/2} dt + \\ + at^{-\alpha/2} \delta(t, \hat{Z}) dt + at^{-\alpha/2} \sum_{r=1}^k \hat{\sigma}_r(t, \hat{Z}) d\xi_r(t), \quad t \geq s, \\ \hat{Y}(s) = s^{\alpha/2} \zeta.$$

Its solution is

$$(10) \quad \hat{Y}(t) = t^{\alpha/2} \exp \{a(1-\alpha)^{-1} B(t^{1-\alpha} - s^{1-\alpha})\} \zeta + \\ + \int_s^t (t/u)^{\alpha/2} \exp \{a(1-\alpha)^{-1} B(t^{1-\alpha} - u^{1-\alpha})\} \cdot$$

$$\cdot [(Q(u) \hat{Z}(u) u^{\alpha/2} - q(u) u^{\alpha/2} + a \delta(u, \hat{Z}) u^{-\alpha/2}) du + au^{-\alpha/2} \sum_{r=1}^k \hat{\sigma}_r(u, \hat{Z}) d\xi_r(u)].$$

Disclosing the brackets, the integral in (10) splits into four ones; the first of them tends to zero in the mean and hence also in probability:

$$(11) \quad E \left| \int_s^t (t/u)^{\alpha/2} \exp \{a(1-\alpha)^{-1} B(t^{1-\alpha} - u^{1-\alpha})\} Q \hat{Z} u^{\alpha/2} du \right| \leq$$

$$\begin{aligned} &\leq \int_s^t (t/u)^{\alpha/2} |\exp\{\cdot\}| |Q| |Z| u^{\alpha/2} du \leq \\ &\leq K \int_s^t \exp\{-\lambda_1(t^{1-\alpha} - u^{1-\alpha})\} \varepsilon(u) u^{-\alpha} du, \quad \lambda_1 > 0, \quad \varepsilon(u) \searrow 0, \end{aligned}$$

where we have utilized the properties of the matrix B and Q ((xiii) and (xii)) and the inequality (9); after the substitution $t^{1-\alpha} - u^{1-\alpha} = v$, the last line of (11) is transformed into

$$K(1-\alpha)^{-1} \int_0^{t^{1-\alpha}-s^{1-\alpha}} e^{-\lambda_1 v} \varepsilon(t(1-v/t^{1-\alpha})^{1/(1-\alpha)}) dv,$$

which tends to 0 for $t \rightarrow \infty$.

The second integral can be written (in view of (xv)) as

$$-\int_s^t (t/u)^{\alpha/2} \exp\{\cdot\} (q_\infty + \varepsilon_1(u)) u^{-\alpha} du, \quad \varepsilon_1(u) \searrow 0;$$

the same substitution changes it into

$$\begin{aligned} &-(1-\alpha)^{-1} \int_0^{t^{1-\alpha}-s^{1-\alpha}} (1-v/t^{1-\alpha})^{-\alpha/(2-2\alpha)} \exp\{a(1-\alpha)^{-1} Bv\} \cdot \\ &\cdot (q_\infty + \varepsilon_1(t(1-v/t^{1-\alpha})^{1/(1-\alpha)})) dv, \end{aligned}$$

which tends for $t \rightarrow \infty$ to

$$\begin{aligned} &-q_\infty(1-\alpha)^{-1} \int_0^\infty \exp\{a(1-\alpha)^{-1} Bv\} dv = -q_\infty a^{-1} \int_0^\infty e^{Bw} dw = \\ &= q_\infty a^{-1} B^{-1}. \end{aligned}$$

The third integral, $a \int_s^t (t/u)^{\alpha/2} \exp\{\cdot\} \delta u^{-\alpha/2} du$, can be again shown to tend to 0 in probability (cf. Lemma 6 in Dupac [2]), as well as the integral

$$a \int_s^t (t/u)^{\alpha/2} \exp\{\cdot\} u^{-\alpha/2} \sum_{r=1}^k (\hat{\sigma}_r(u, Z) - s_r) d\xi_r(u)$$

(cf. the same paper, formulas (13), (14)).

As the first term in (10), $t^{\alpha/2} \exp\{\cdot\} \zeta$, tends obviously to 0 owing to the properties of B , we get thus that the distribution of $\hat{Y}(t) - q_\infty a^{-1} B^{-1}$ is asymptotically equivalent to the distribution of

$$a \int_s^t (t/u)^{\alpha/2} \exp\{a(1-\alpha)^{-1} B(t^{1-\alpha} - u^{1-\alpha})\} u^{-\alpha/2} \sum_{r=1}^k s_r d\xi_r(u),$$

which is, however, a Gaussian process with zero mean and a covariance matrix, which can be calculated in a straightforward way, using the same substitution as above, and shown to tend to $a \int_0^{\infty} e^{Bv} S e^{B^T v} dv$, for $t \rightarrow \infty$. The rest of the proof consists in proving the asymptotic equivalence of distributions of

$$t^{z/2}(X^x(t) - \theta(t)) = t^{z/2} Z^z(t) \quad \text{and of} \quad \hat{Y}(t) = t^{z/2} Z^z(t)$$

for properly related z and ζ ; it is exactly the same as the end of the proof of the Theorem in Dupač [2].

It should be pointed out, that the proofs in the present paper as well as in the paper Dupač [2] more or less follow the pattern of proofs in Nevel'son, Has'minskij [4], Chapt. 6.

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