

OPTIMAL CONTROL PROBLEM FOR DIFFUSION PROCESSES WITH LONG RUN AVERAGE COST PER UNIT TIMES

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1. INTRODUCTION

In this paper, we study the control problem for dynamical systems whose state X_t at time t is described by a stochastic differential equation of the form

$$\begin{cases} dX_t = b(X_t, U_t) dt + \sigma(X_t) dW_t \\ X_0 = x \in R^d \end{cases} \quad (1.1)$$

where (W_t) is a Wiener process defined on a stochastic basis $(\Omega, \mathcal{F}_t, P)$, satisfying the standard conditions and b, σ are given functions.

As usually the control process (U_t) is supposed to be progressively measurable with respect to the filtration \mathcal{F}_t , with value in a metric compact space A . For any admissible control (U_t) we consider the solution $X_t^U(x)$ of (1.1) associated with the control (U_t) starting from x at $t = 0$. We consider the average cost of (U_t) per unit time

$$J(x, U) = \limsup_{T \rightarrow \infty} \frac{1}{T} E \int_0^T h(X_t, U_t) dt$$

which to minimise the cost $J(x, U)$ over the class U_{ad}

This problem has been dealt with in various ways. In [7] admissible controls are defined as maps from R^d into A satisfying the Lipschitz condition and under this assumption the author has shown that there exists a Lipschitz feedback control. Kushner in [6] used relaxed control for problems for "wideband noise driven" which are "close" to a diffusion. In this article we use methods dealt with by [5] for discrete cases to develop these results. Under the elliptic and ergodic hypotheses we show that there exists an optimal Markovian measure. Hence using the ergodic theorem we can prove the existence of an optimal Markovian control process.

2. NOTATIONS AND HYPOTHESES

Hypotheses

Let A be a metric compact space, called the actions space and let

$$\begin{aligned} \sigma &: R^d \rightarrow d \times d \text{ - matrices} \\ b &: R^d \times A \rightarrow R^d \\ h &: R^d \times A \rightarrow R \end{aligned} \quad (2.1)$$

Throughout this paper we suppose that the following hypotheses are satisfied:

Hypothesis 1: σ, b, h are continuous functions in (x, a) and h is bounded; σ locally Lipschitz continuous.

We consider the generator associated with the coefficients σ and b of the equation (1) by

$$L^a f(x) = \frac{1}{2} \sum a_{ij} \frac{\partial^2 f}{\partial x_i \partial x_j} + \sum b_i(x, a) \frac{\partial f}{\partial x_i}$$

where a_{ij} is the ij -th element of the matrix $\sigma\sigma^*(x)$ and $b(x) = (b_1, b_2, \dots, b^d)^T$; $a \in A$. operator L defined on the space of twice continuous differential function with bounded derivatives, say $C_b^2(R^d)$

Hypothesis 2: The operator L^a is uniformly elliptic. That is: there is a $\alpha > 0$ such that

$$\sigma\sigma^*(x) \geq \alpha I \quad \text{for any } x \in R^d$$

2.2. Admissible control

We define admissible controls as the case of finite horizon. Let U be an admissible control then X_t^U is the solution of (1) associated with U if and only if

$$C_t(f, U) = f(X_t) - f(X_s) - \int_s^t L^{U_h} f(X_h) dh$$

is a P -martingale after s for any $f \in C_b^2(R^d)$. Therefore we have the following definition

Definition 1. An admissible control is a term $U^0 = (\Omega, \mathcal{F}_t, P, X_t, U_t, \nu)$ such that:

- (i) $(\Omega, \mathcal{F}_t, P)$ is a probability space with the filtration \mathcal{F}_t .
- (ii) U_t is a progressively measurable process with values in A .
- (iii) (X_t) is a continuous process with values in R^d such that

$$C_t(f, U) = f(X_t) - f(X_s) - \int_0^t L^{U_h} f(X_h) dh$$

is a P -martingale for any $f \in C_b^2(R^d)$

- (iv). X_0 has the distribution ν .

The set of admissible controls starting from ν at $t = 0$ is denoted by \mathcal{A}_0 . Let $U^0 \in \mathcal{A}_0$ consider the payoff

$$J_0(\nu, U^0) = \limsup_{T \rightarrow \infty} \frac{1}{T} P \int_0^T h(X_t, U_t) dt$$

where $P(\Phi) = \int \Phi dP$. Our aim is to minimise $J_0(\nu, U)$ over class $\mathcal{A}^0(\nu)$

2.3. Relaxed controls

We write for V the space of generalised actions consisting of random measures on $R^+ \times R^d$ of the form $dt.q(t, da)$ where dt is the Lebesgue measure on R^+ and $q(t, da)$ is a probability

any $t \in R^+$. With the vague topology, V is a metric compact space. This is the same as the topology on $R^+ \times A$ (see [3]).

We write, from now on, $f(t, q_t)$ for $\int_A f(t, a) q_t(da)$ for any measurable f . The space V is endowed with its Borel σ -field \mathcal{V} which is also the smallest σ -field such that the maps $\int_{R^+} \int_A f(s, a) q(ds, da)$ are measurable for any measurable function f continuous in a with compact support. We also introduce the filtration (\mathcal{V}_t) where \mathcal{V}_t is the σ -algebra generated by measures $\{1_{[0,t]} q : q \in \mathcal{V}\}$. The details of these definitions can be referred in [1].

Definition 2. A relaxed control is a term $U = (\Omega, \mathcal{F}_t, P, X_t, q, \nu)$ such that

- (i) $(\Omega, \mathcal{F}_t, P)$ is a probability space with the filtration \mathcal{F}_t satisfying the general hypotheses.
- (ii) q is a V -valued stochastic process, \mathcal{F}_t -adapted.
- (iii) (X_t) is a continuous process with values in R^d such that

$$C_t(f, U) = f(X_t) - f(X_0) - \int_0^t \int_A L^a f(X_h) q(h, da) dh \quad (2.5)$$

P -martingale for any $f \in C_b^2(R^d)$

- (iv) The distribution of X_0 is ν .

The set of relaxed control with the initial distribution ν is denoted by $\mathcal{A}(\nu)$. For any $U \in \mathcal{A}(\nu)$ consider the payoff

$$J(\nu, U) = \limsup_{T \rightarrow \infty} \frac{1}{T} P \int_0^T h(t, a) q(t, da) ds$$

$$J(\nu) = \inf \{ J(\nu, U) : U \in \mathcal{A}(\nu) \}; \quad J^* = \inf_{\nu} J(\nu)$$

write $J(x)$ for $J(\nu)$ whenever ν is Dirac mass at x . The pair (ν^*, U^*) where $U^* \in \mathcal{A}(\nu^*)$ is said to be minimum if $J(\nu^*, U^*) = J^*$. The term $(\Omega, \mathcal{F}_t, P, X_t, q)$ is called optimal if for any $x \in R^d$ relaxed control $U = (\Omega, \mathcal{F}_t, P, X_t, q, x)$ satisfies the relation

$$J(x) = J(x, U)$$

the following we can see that the set of admissible controls $\mathcal{A}_0(\nu)$ is complete in the sense that

$$J^* = \inf_{\nu} \inf \{ J(\nu, U) : U \in \mathcal{A}(\nu) \} = \inf_{\nu} \inf \{ J_0(\nu, U^0) : U^0 \in \mathcal{A}_0 \}$$

Rules of Controls

Similar as in [1], we formulate the problem on the canonical space \mathcal{X} consisting of continuous functions from R^+ into R^d endowed with the topology of uniform convergence on every bounded interval. Let \mathcal{X}_t be the natural right-continuous filtration and (X_t) be the canonical process defined on \mathcal{X} . We put

$$\bar{\mathcal{X}} = \mathcal{X} \times V; \quad \bar{\mathcal{X}}_t = \bigcap_{s > t} \mathcal{X}_s \oplus \mathcal{V}_s$$

Definition 3. A control rule (more briefly: a rule) is a probability measure R on $\bar{\mathcal{X}}$ such that system $(\bar{\mathcal{X}}, \bar{\mathcal{X}}_t, R, X_t, q, \nu)$ is a relaxed control.

The set of rules starting from ν is denoted by $\mathcal{R}(\nu)$. In term of control rules, optimal relaxed control can be interpreted as a family of rules $\{R_x \in \mathcal{R}(x) : x \in R^d \text{ such that } J(x, R_x) = J(x) \text{ any } x \in R^d\}$.

We are now able to define markovian controls by two ways : either we consider them in term $U = (\Omega, \mathcal{F}_t, P_x, X_t, q, x)$ such that $q = dt \times q(X_t, da)$ or as a family of rules $\{R_x : x \in R^d\}$ such that the term $(\bar{X}, \bar{X}_t, X_t, R_x, x)$ is a family of homogeneous markovian process . We will show that two these definitions are equivalent.

Theorem 1. *The family of rules $\{R_x : x \in R^d\}$ is markovian iff there exists a valued - measure map q from R^d into $\mathcal{P}(A)$ and a relaxed control $(\Omega, \mathcal{F}_t, P, X_t, dt q(X_t, da))$ such that R_x is the optimal rule of the couple $(X_t, dt \times q(t, da))$ under P given $X_0 = x$.*

Proof. The sufficient condition is obvious. The necessary one is proved by a similar way as Theorem 6.7 in [1] and we do not reproduce it here. \diamond

In order to show the existence of optimal rules we follow the argument dealt with by Kurikawa in [5] for the discrete case. First, we prove that there exists an optimal pair (ν^*, R^*) such that ν^* is stationary distribution of X_t under R . After that by virtue of recurrent property of (X_t) we show that this rule is an optimal markovian control .

Let $U \in \mathcal{A}(\nu)$ be a relaxed control we put

$$\gamma_T^U(D) = \frac{1}{T} P \int_0^T 1_D(X_t) dt \quad D \in \mathcal{B}(R^d), \quad T > 0.$$

Hypothesis 3. *For any ν , the family*

$$\{\gamma_T^U(\cdot) : T > 0, U \in \mathcal{A}(\nu)\} \quad (3)$$

is tight.

Hypothesis 3 will be true if we consider reflected diffusion processes in a bounded domain R^d (see [7]) or if the following hypothesis 3' is satisfied. Set

$$\Phi(x) = \sum_{i,j=1}^d a_{ij}(x) x_i x_j; \quad \Psi^a(x) = \frac{1}{\Phi(x)} \left[\frac{1}{2} \sum_{ij} a_{ij}(x) + \sum_{i=1}^d b_i(x, a) x_i \right]$$

and

$$\Phi^+(r) = \sup_{|x|=\sqrt{2r}} \Phi(x); \quad \Psi^+(r) = \sup_{|x|=\sqrt{2r}} \sup_{a \in A} \Psi^a(x)$$

It is easy to check that $\Phi^+ > 0$ is locally Lipschitz continuous and Ψ^+ is measurable (see [2] 374). Hence there exists the minimal diffusion process ξ_t generated by the operator

$$L^+ = \Phi^+(r) \left[\frac{d^2}{dr^2} + \Psi^+(r) \cdot \frac{d}{dr} \right]$$

Hypothesis 3': *The family $\{\sigma_T^\xi(\cdot); T > 0\}$ where*

$$\sigma_t^\xi(B) = \frac{1}{T} \int_0^T P(\xi_t \in B); \quad T > 0$$

it.

Indeed, let $(\Omega, \mathcal{F}_t, P, X_t, q)$ be an arbitrary relaxed control. We put $H(t) = \int_0^t \frac{\Phi(X_s)}{\Phi + (|X_s|)} ds$ then easy to verify that $\frac{\lambda}{\Lambda} t \leq H(t) \leq t$ where λ and Λ respectively is the minimum of smallest maximum of biggest eigenvalues of the matrix $\sigma\sigma^*(x)$ in x . Moreover, $H(t)$ is a increasing function. We write for $G(t)$ its inverse function. By the comparison theorems (see [2], pp 371; we have

$$|X(G(t))|^2 \leq \xi_t \quad \text{a.s}$$

on the other hand

$$\begin{aligned} \int_0^T 1_{\{\xi_t > c\}} dt &\geq \int_0^T 1_{\{|X(t)|^2 > c\}} dt = \int_0^{G(T)} 1_{\{|X(u)|^2 > c\}} H'(u) du \\ &\geq \frac{\lambda}{\Lambda} \int_0^{G(T)} 1_{\{|X(u)|^2 > c\}} du \end{aligned}$$

implies that

$$\frac{1}{T} \int_0^T 1_{\{\xi_t > c\}} dt \geq \int_0^{\frac{\Lambda}{\lambda}} 1_{\{|X(u)|^2 > c\}} dt$$

by hypothesis 3', $\{\gamma_T^U(\cdot), T > 0\}$ is tight.

Lemma 2. (see [5] Lemma 2.1). For any initial distribution ν and for any $R \in \mathcal{R}(\nu)$, there is a probability measure μ on $R^d \times A$ such that

$$\int h(x, a) \mu(dx, da) \leq J(\nu, R) \quad (2.7)$$

$$\int_{R^d} \int_A L^\alpha f(x) \mu(dx, da) = 0 \quad (2.8)$$

any $f \in C_b^2$

f. Let ν and $R \in \mathcal{R}(\nu)$ arbitrary. We put

$$\gamma_T^R(D) = \frac{1}{T} R \int_0^T 1_D(X_t, q_t) dt; \quad D \in \mathcal{B}(R^d \times A)$$

$\{\gamma_T^R : T > 0\}$ is tight then there exists a sequence $(T_n) \rightarrow \infty$ such that

$$\gamma_{T_n}^R(\cdot) \xrightarrow{\text{weak}} \mu(\cdot)$$

if h is a continuous function then

$$\int h(x, a) \mu(dx, da) = \lim_{n \rightarrow \infty} \int h(x, a) \gamma_{T_n}^R(dx, da) \leq \limsup \frac{1}{T} R \int_0^T h(X_t, q_t)$$

on the other hand, for any $f \in C_b^2$ we have

$$C_t(f, U) = f(X_t) - f(X_s) - \int_s^t L^{U_h} f(X_h) dh$$

P -martingale then by the law of large numbers it follows that

$$0 = \lim_{T_n} \frac{1}{T_n} E \int_0^{T_n} Lf(X_t, q_t) dt = \int L^\alpha f(x) \mu(dx, da)$$

so we get (2.8) \diamond

Lemma 2. *If $\{\mu_n\}$ is a sequence of probability measures satisfying Condition (2.8) then $\{\mu_n\}$ is tight.*

Proof. Since A is compact then the sequence of projections of $\{\mu_n\}$ on A is automatically compact. Suppose that $\mu_n(dx, da) = \mu_n(dx)\nu_n(x, da)$ and $\bar{b}_n(x) = \int b(x, a)\nu_n(x, da)$. Let X_n be the solution of the equation

$$\begin{aligned} dX_n(t) &= \bar{b}_n(X_n(t)) dt + \sigma(X_n(t))dW_t \\ X_n(0) &= 0 \in R^d \end{aligned} \quad (2.9)$$

Then $U^n = (\Omega, \mathcal{F}_t, P, X_n(t), \nu_n(X_n, da), 0)$ is an relaxed control and $U^n \in \mathcal{A}(\delta_0)$. On the other hand, from (2.3) and (2.8) it follows that μ_n is the unique invariant measure of Equation (2.9) :

$$\mu_n(D) = \lim_{T \rightarrow \infty} \frac{1}{T} P \int_0^T 1_D(X_n(t)) dt$$

for any set D good enough. By Hypothesis 3, the sequence $\{\gamma_T^{U^n}(\cdot) = \frac{1}{T} P \int_0^T 1_D(X_n(t)) dt; 0; n > 0\}$ is tight then it follows the tightness of $\{\mu_n\}$. The proof is complete. \diamond

Lemma 3: *there exists a probability measure ν^* and a markovian control rule R^* such that*

$$J(\nu^*, R^*) = J^*$$

i.e., the pair (ν^, R^*) is optimal.*

Proof. Let $\{\nu_n, R_n\}$ be a sequence minimising $J(\cdot, \cdot)$. This means that

$$\lim_{n \rightarrow \infty} J(\nu_n, R_n) = J^*$$

By Lemma 2, for any n there exists a probability measure on $R^d \times A$, namely $\mu_n(\cdot)$, such that

$$\int h(x, a) \mu_n(dx, da) \leq J(\nu_n, R_n)$$

From Lemma 2, the sequence $\{\mu_n\}$ is tight. Then there is a probability measure μ on $R^d \times A$ and a sequence $(n_k) \rightarrow \infty$ such that

$$\mu_{n_k} \xrightarrow{\text{weak}} \mu$$

Hence, we have

$$\int h(x, a) \mu(dx, da) = \lim_{n_k \rightarrow \infty} \int h(x, a) \mu_{n_k}(dx, da) \leq \liminf_{n_k \rightarrow \infty} J(\nu_{n_k}, R_{n_k}) = J^*$$

and

$$\int L^a f(x) \mu(dx, da) = 0$$

Suppose that $\mu(dx, da) = \nu(dx) \cdot q(x, da)$. We put

$$b^q(x) = \int b(x, a) q(x, da)$$

the equation

$$dX_t = b^q(X_t) dt + \sigma(X_t) dW_t \quad (2.10)$$

is a unique stationary solution with the stationary distribution $\nu(dx)$. It is easy to check that the pair (ν, R_ν) is optimal where R_ν is the law of $(X_t, dt \cdot q(X_t, da))$ on \bar{X} . The result follows. \diamond

Since the diffusion matrix $a(x) = \sigma(x)\sigma^*(x)$ is non-degenerate then the invariant measure ν is absolutely continuous with respect to the Lebesgue measure on R^d and support $\nu = R^d$ (see [AK]) in the limit

$$\lim_{T \rightarrow \infty} \frac{1}{T} E \int_0^T \int_A h(X_t, a) q(X_t, da) dt = \int h(x, a) q(x, da) \nu(dx) = J^*$$

for any $x \in R^d$ a.s (see [7] Th. 6.1). This means that $q(x, da)$ is a markovian optimal control.

Denoting $A(x)f = \frac{1}{2} \sum a_{ij} \frac{\partial^2 f}{\partial x_i \partial x_j}$ and $\langle b^a, \nabla f \rangle = \sum b_i(x, a) \frac{\partial f}{\partial x_i}$, we now consider the equation

$$A(x)\Phi + \langle b^q, \nabla \Phi \rangle + h^q(x) = J^* \quad (2.11)$$

equation has a generalised solution Φ since σ is non degenerate. Let

$$s(x) = \{a \in A : A(x)\Phi + \langle b^a(x), \nabla \Phi \rangle + h(x, a) \leq J^*\}$$

2.10) $s(x) \neq \emptyset$ and $q(x, s(x)) \neq 0$ for any $x \in R^d$. Hence there is a measurable selection $R^d \rightarrow A$ such that

$$\frac{1}{2} \sum a_{ij} \frac{\partial^2 \Phi}{\partial x_i \partial x_j} + \sum b_i(x, u^*(x)) \frac{\partial \Phi}{\partial x_i} + h(x, u^*(x)) \leq J^*$$

in virtue of the generalized Ito's formula (see [4]) we have

$$\begin{aligned} E\Phi(X_t) &= E\Phi(X_0) + E \int_0^t (A(x)\Phi + \langle b^{u^*}, \nabla \Phi \rangle) ds \\ &\leq E\Phi(X_0) + E \int_0^t [J^* - h(X_s, u^*(X_s))] ds \end{aligned}$$

inequality implies that

$$\limsup \frac{1}{T} E \int_0^T h(X_t, u(X_t)) dt \leq J^*$$

means that u^* is an optimal control process.

Lemma 3: Under Hypotheses 1, 2, 3, there exists an optimal control process u^* which satisfies the equation

$$A(x)\Phi + \langle b^{u^*}, \nabla \Phi \rangle + h(x, u^*(x)) \leq J^*$$

From this result we can follow that the set \mathcal{A}^0 is dense in \mathcal{A} as we have mentioned. The proof is so we omit it here.

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ĐIỀU KHIỂN TỐI ƯU CÁC QUÁ TRÌNH KHUẾCH TÁN VỚI GIÁ TRUNG BÌNH THEO THỜI GIAN

Nguyễn Hữu Dư

Trường ĐH Khoa học tự nhiên - ĐHQG Hà Nội

Bài báo đề cập đến bài toán điều khiển tối ưu các quá trình khuếch tán với giá trung bình theo thời gian. Dưới giả thiết về tính compact tương đối của lớp các điều khiển và sự không s biến của hệ số khuếch tán của quá trình tín hiệu, nhờ phương pháp tương tự như trong [5], chúng tôi chỉ ra sự tồn tại của điều khiển tối ưu Markov. Bài báo là sự mở rộng phương pháp trong từ rời rạc lên trường hợp liên tục.