

ON STOCHASTIC RELAXED CONTROL FOR PARTIALLY OBSERVED DIFFUSIONS

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§1. Introduction

In this paper we are concerned with stochastic relaxed control problems of the following kind. Let $X(t)$, $t \geq 0$, denote the state of a process being controlled, $Y(t)$, $t \geq 0$, the observation process and $p(t, \cdot)$ a relaxed control, that is a process with values probability measures on the control region Γ . The state and observation processes are governed by stochastic differential equations

$$(1.1) \quad \begin{cases} dX(t) = \alpha(X(t))dB(t) + \int_{\Gamma} r(X(t), u)p(t, du)dt \\ X(0) = \xi \end{cases}$$

and

$$(1.2) \quad \begin{cases} dY(t) = h(X(t))dt + dW(t) \\ Y(0) = 0 \end{cases}$$

where B and W are independent Brownian motions with values in R^n and R^m respectively, (put $m = 1$ for simplicity).

The problem is to maximize a criterion of the form

$$J = Ef(X(T))$$

by a suitable choice of admissible relaxed control p . In a customary version of stochastic control under partial observation, $p(t, \cdot)$ is measurable with respect to σ -field generated by the observation process $Y(s)$, $s \leq t$. Instead of discussing the problem of this type, we treat some wider class of admissible relaxed controls (see §2), inspired by Fleming-Pardoux [8]. Roughly speaking, our problem is the following; Let

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$$(1.3) \quad L(T) = \exp \left\{ \int_0^T h(X(s)) dY(s) - \frac{1}{2} \int_0^T |h(X(s))|^2 ds \right\}.$$

Then B and Y turn out as independent Brownian motions under a new probability \dot{P} , defined by

$$(1.4) \quad \frac{d\dot{P}}{dP} = L^{-1}(T)$$

appealing to the so-called Girsanov transformation. For admissibility we merely require that $p(t, \cdot)$ is independent of future increments of $Y(\theta) - Y(s)$, $\theta, s \geq t$, and B , with respect to \dot{P} . Moreover we are concerned with $q(dt, du)$ instead of $p(t, du)dt$. (see Definition 1). Thus the criterion J can be expressed as

$$(1.5) \quad J = \dot{E}f(X(T))L(T)$$

where \dot{E} stands for the expectation with respect to \dot{P} , and $X(t)$ is a solution of the following system equation;

$$(1.6) \quad dX(t) = \alpha(X(t))dB(t) + \int_r \gamma(X(t), u)q(dt, du).$$

Under Lipschitz continuity and boundedness of α and γ , (1.6) has a unique solution (Theorem 1).

In Section 2 we introduce some metric spaces which are appropriate to our optimization problems. In Section 3 we prove the compactness of spaces of solutions and relaxed controls q . This guarantees the existence of optimal one (Theorem 3).

In the latter half we treat a nonlinear semigroup associated with relaxed control under partial observation. In this case we regard, as the state space, the unnormalized conditional distribution $\Lambda(t)$ of $X(t)$ given past observation and control. Hence $\Lambda(t)$ is a process valued in measures on R^n and satisfying the Zakai equation. Thus our problems turn out as optimization problems of measure valued processes. After we prove the continuity of $\Lambda(t)$ with respect to initial distribution $X(0)$ and data of past observation and control, (see Theorem 5), we construct a nonlinear semigroup $S(t)$, $t \geq 0$, on a Banach lattice of bounded and uniformly continuous functions, defined on the space of measures (Theorem 7). Following Fleming [6], we show that the generator of $S(t)$ relates to a dynamic programming equation, so-called Mortensen's equation.

§ 2. Notations and preliminaries

Let (Ω, F, \dot{P}) be a probability space. Let B and Y be n -dimensional and 1-dimensional Brownian motions, defined on (Ω, F, \dot{P}) respectively. Γ is a given convex compact subset of R^k , called a control region. $M(\Gamma)$ denotes the totality of positive finite measures defined on $\mathcal{B}(\Gamma)$ (= Borel field of Γ). By $\hat{M}([0, T] \times \Gamma)$ we denote the set of all mappings $\lambda: [0, T] \times \mathcal{B}(\Gamma) \rightarrow [0, T]$ such that

- 0) $\lambda(0, A) = 0 \quad \forall A \in \mathcal{B}(\Gamma)$
- i) $\lambda(t, \Gamma) = t \quad \forall t \geq 0$
- ii) $\lambda(t, \cdot) \in M(\Gamma)$ for all $t > 0$
- iii) $\lambda(t, A)$ is increasing in t for all $A \in \mathcal{B}(\Gamma)$
- iv) $\sup_{A \in \mathcal{B}(\Gamma)} |\lambda(s, A) - \lambda(t, A)| = |t - s|$.

From (ii) and (iii), λ determines a measure on $[0, T] \times \Gamma$ and $\lambda([s, t], \cdot) \equiv \lambda(t, \cdot) - \lambda(s, \cdot) \in M(\Gamma)$, if $t > s$, and $\lambda([s, t], \Gamma) = t - s$.

Let q be a mapping; $[0, T] \times \mathcal{B}(\Gamma) \times \Omega \rightarrow [0, 1]$, such that

- v) for all $A \in \mathcal{B}(\Gamma)$, $q(\cdot, A, \cdot)$ is $\mathcal{B}[0, T] \times F$ -measurable
- vi) $q \in \hat{M}([0, T] \times \Gamma)$ \dot{P} -almost surely.

DEFINITION 1. $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, Y, q)$ is called an admissible (relaxed) system, if ξ is an n -random vector on (Ω, F, \dot{P}) , which is independent of (B, Y, q) , B and (Y, q) are independent and the increments $(Y(t) - Y(s), t \geq s)$ are independent of $\sigma_s(Y, q)$ (= σ -field generated by $Y(\theta), \theta \leq s$ and $q(\theta, A), \theta \leq s, A \in \mathcal{B}(\Gamma)$).

DEFINITION 2. The component q of \mathcal{A} is called a relaxed control, and we denote it by $q_{\mathcal{A}}$ when \mathcal{A} is stressed. $\lambda \in \hat{M}([0, T] \times \Gamma)$ can be regarded as a relaxed control. \mathfrak{A} denotes the totality of admissible systems.

Let $\alpha(x)$ be a symmetric $n \times n$ matrix valued function on R^n and γ an n -vector continuous function on $R^n \times \Gamma$. We assume the following conditions

- (A1) $|g(x, u)| \leq b, \quad \forall x \in R^n, u \in \Gamma, g = \alpha, \gamma$
- (A2) $|g(x, u) - g(x', u)| \leq K|x - x'|, \quad \forall x, x' \in R^n, u \in \Gamma, g = \alpha, \gamma$.

For an admissible system $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, Y, q)$ we consider the stochastic differential equation (SDE in short)

$$(2.1) \quad \begin{cases} dX(t) = \alpha(X(t))dB(t) + \int_{\Gamma} \gamma(X(t), u)q(dt, du) \\ X(0) = \xi. \end{cases}$$

THEOREM 1. *There exists a unique solution X of (2.1) which is $\sigma_t(\xi, B, q)$ -progressively measurable and has continuous paths.*

Proof. We apply a usual successive approximation method. We define X_n in the following way

$$(2.2) \quad \begin{aligned} X_0(t) &= \xi \\ X_{n+1}(t) &= \xi + \int_0^t \alpha(X_n(s)) dB(s) + \int_0^t \int_{\Gamma} \gamma(X_n(s), u) q(ds, du) \\ & \qquad \qquad \qquad n = 0, 1, 2, \dots \end{aligned}$$

Then, X_n is $\sigma_t(\xi, B, q)$ -progressively measurable and has continuous paths by (iv) and (A1).

$$\begin{aligned} X_{n+1}(t) - X_n(t) &= \int_0^t (\alpha(X_n(s)) - \alpha(X_{n-1}(s))) dB(s) \\ & \quad + \int_0^t \int_{\Gamma} (\gamma(X_n(s), u) - \gamma(X_{n-1}(s), u)) q(ds, du) \end{aligned}$$

So, using (A1) and (A2) we see

$$\begin{aligned} & \left(\int_0^t \int_{\Gamma} |\gamma(X_n(s), u) - \gamma(X_{n-1}(s), u)| q(ds, du) \right)^2 \\ & \leq \int_0^t \int_{\Gamma} |\gamma(X_n(s), u) - \gamma(X_{n-1}(s), u)|^2 q(ds, du) q(t, \Gamma) \\ & \leq Kt \int_0^t \int_{\Gamma} |X_n(s) - X_{n-1}(s)|^2 q(ds, du) \\ & = Kt \int_0^t |X_n(s) - X_{n-1}(s)|^2 ds. \end{aligned}$$

Putting $\rho_n(t) = E|X_{n+1}(t) - X_n(t)|^2$, we have

$$\rho_n(t) \leq K_1 \int_0^t \rho_{n-1}(s) ds, \quad \text{for } \forall t \leq T,$$

with some $K_1 = K_1(T)$. This implies

$$(2.3) \quad \rho_n(t) \leq \frac{t^{n-1} K_1^{n-1}}{(n-1)!} \dot{E}|\xi|^2.$$

Therefore

$$\sum_n \dot{E}|X_{n+1}(t) - X_n(t)| \leq \sum \sqrt{\rho_n(t)} < \infty.$$

This implies that $X_n(t)$ converges \dot{P} -almost surely. Hence $X(t) = \lim_{n \rightarrow \infty} X_n(t)$

can be regarded as $\sigma_t(\xi, B, q)$ -progressively measurable and moreover a martingale inequality tells us that, as $n \rightarrow \infty$,

$$(2.4) \quad \int_0^t \alpha(X_n(s)) dB(s) \rightarrow \int_0^t \alpha(X(s)) dB(s) \quad \text{uniformly in } t \in [0, T],$$

\dot{P} -almost surely. On the other hand

$$\begin{aligned} & \sup_{t \leq T} \left| \int_0^t \int_R \gamma(X_n(s), u) - \gamma(X(s), u) q(ds, du) \right. \\ & \leq \int_0^T \int_R |\gamma(X_n(s), u) - \gamma(X(s), u)| q(ds, du) \\ & \leq K \int_0^T \min(|X_n(s) - X(s)|, 2b) ds. \end{aligned}$$

By virtue of the convergence theorem we get, as $n \rightarrow \infty$,

$$(2.5) \quad \int_0^t \int_R \gamma(X_n(s), u) q(ds, du) \rightarrow \int_0^t \int_R \gamma(X(s), u) q(ds, du),$$

uniformly in $t \in [0, T]$, \dot{P} -almost surely.

Combining (2.4) and (2.5) with (2.2), X turns out as a solution of (2.1) and $X_n(t)$ converges to $X(t)$ uniformly in $t \in [0, T]$ \dot{P} -almost surely. Hence X has continuous paths.

Let Y be a $\sigma_t(\xi, B, q)$ -progressively measurable solution of (2.1). Then, applying a routine method, we can easily see for $\forall t$,

$$X(t) = Y(t) \quad \dot{P}\text{-almost surely.}$$

This completes the proof of Proposition 2.1.

L denotes the Prohorov metric for probability measures. That is following, [11]. Let ε_{21} be the infimum of ε such that

$$\mu_1(F) \leq \mu_2(U_\varepsilon(F)) + \varepsilon \quad \text{for all closed subset } F,$$

where $U_\varepsilon(F)$ is the ε -neighbourhood of F . ε_{12} is defined by switching μ_1 and μ_2 . Set

$$(2.6) \quad L(\mu_1, \mu_2) = \max(\varepsilon_{12}, \varepsilon_{21}).$$

Put

$$(2.7) \quad M(\Gamma, t) = \{\lambda \in M(\Gamma); \lambda(\Gamma) = t\}, \quad t > 0.$$

Define a metric ρ_t as follows

$$(2.8) \quad \rho_t(\lambda_1, \lambda_2) = L\left(\frac{\lambda_1}{t}, \frac{\lambda_2}{t}\right), \quad \lambda_i \in M(\Gamma, t).$$

Since Γ is compact, $(M(\Gamma, t), \rho_i)$ is a compact metric space. Put $D = \{r_i, \text{rational } \in [0, T], i = 1, 2, \dots\}$ and

$$(2.9) \quad \tilde{M}_T = \prod_{i=1}^{\infty} M(\Gamma, r_i).$$

We endow \tilde{M}_T with a metric \tilde{d}_T such that

$$(2.10) \quad \tilde{d}_T(\tilde{\lambda}, \tilde{\mu}) = \sum_{n=1}^{\infty} \frac{1}{2^n} \rho_{r_n}(\lambda_n, \mu_n)$$

where $\tilde{\lambda} = (\lambda_1, \lambda_2, \dots)$ and $\tilde{\mu} = (\mu_1, \mu_2, \dots)$. Hence $\tilde{\lambda}_k, k = 1, 2, \dots$ is a \tilde{d}_T -Cauchy sequence, iff $\lambda_{k,i}, k = 1, 2, \dots$ is a ρ_{r_i} -Cauchy sequence for any component i . Therefore again $(\tilde{M}_T, \tilde{d}_T)$ is a compact metric space.

Since $\lambda \in \tilde{M}([0, T] \times \Gamma)$ is determined by $\tilde{\lambda} = (\lambda(r_1), \lambda(r_2), \dots) \in \tilde{M}_T$, we define a metric \hat{d}_T on $\hat{M}([0, T] \times \Gamma)$ by

$$(2.11) \quad \hat{d}_T(\lambda, \mu) = \tilde{d}_T(\tilde{\lambda}, \tilde{\mu}).$$

PROPOSITION 2.1. $\hat{M}([0, T] \times \Gamma, \hat{d}_T)$ is a compact metric space.

Proof. Let $\lambda_k(r_i, \cdot)$ converge to $\lambda_{(i)}$ in ρ_{r_i} . Then $\lambda_{(i)} \in M(\Gamma, r_i)$ and for $g \in C_b(\Gamma)$ (=bounded continuous function on Γ).

$$(2.12) \quad \int_{\Gamma} g(u) \lambda_k(r_i, du) \rightarrow \int g(u) \lambda_{(i)}(du), \quad \text{as } k \rightarrow \infty.$$

Define $\lambda(r_i, A)$ by $\lambda(r_i, A) = \lambda_{(i)}(A)$. Then putting $g = 1$ in (2.12), we see

$$\lambda(r_i, \Gamma) = r_i.$$

Let $r_i > r_j$ and set $R(A) = \lambda(r_i, A) - \lambda(r_j, A)$. Then R is a signed measure on Γ . Since $\lambda_k(r_i, \cdot) - \lambda_k(r_j, \cdot) \in M(\Gamma, r_i - r_j)$ and $\lambda_k(r_i, \cdot) - \lambda_k(r_j, \cdot)$ converges to R weakly by (2.12). R turns out as a positive measure and $R \in M(\Gamma, r_i - r_j)$. This means

$$(2.13) \quad \lambda(r, A) \text{ is increasing in rational } r$$

and

$$(2.14) \quad |\lambda(r_i, A) - \lambda(r_j, A)| \leq |\lambda(r_i, \Gamma) - \lambda(r_j, \Gamma)| = r_i - r_j.$$

Now we will construct λ which corresponds to $(\lambda(r_i, \cdot), i = 1, 2, \dots) \in \tilde{M}_T$, in the following way,

$$(2.15) \quad \lambda(t, A) = \lim_{r_i \uparrow t} \lambda(r_i, A).$$

Then λ clearly satisfies the conditions (0) ~ (iv), namely $\lambda \in M([0, T] \times \Gamma)$ and $\hat{d}_T(\lambda_k, \lambda) \rightarrow 0$, as $k \rightarrow \infty$.

Remark. $\rho_t(\lambda_k(t), \lambda(t)) \leq (4/r_t)|t - r_t| + \rho_{r_t}(\lambda_k(r_t), \lambda(r_t))$ by condition (iv). Hence

$$\lambda_k(t) \rightarrow \lambda(t) \quad \text{in } \rho_t.$$

For $g \in C([0, T] \times \Gamma)$ (= continuous function on $[0, T] \times \Gamma$)

$$(2.16) \quad \int_0^t g(s, u) \lambda_k(ds, du) \rightarrow \int_0^t g(s, u) \lambda(ds, du), \quad \text{as } k \rightarrow \infty.$$

Proof of (2.16). Since g is uniformly continuous, $g([2^n t]/2^n, u)$ converges to $g(t, u)$ uniformly on $[0, T] \times \Gamma$, where $[c]$ is the largest integer $\leq c$. Suppose $\sup_{s, u} |g([2^n s]/2^n, u) - g(s, u)| < \varepsilon$. Then

$$\begin{aligned} & \left| \int_0^t \int_{\Gamma} g(s, u) \lambda_k(ds, du) - \int_0^t \int_{\Gamma} g(s, u) \lambda(ds, du) \right| \\ & \leq \int_0^t \int_{\Gamma} \left| g(s, u) - g\left(\frac{[2^n s]}{2^n}, u\right) \right| (\lambda_k(ds, du) + \lambda(ds, du)) \\ & \quad + \left| \int_0^t \int_{\Gamma} g\left(\frac{[2^n s]}{2^n}, u\right) \lambda_k(ds, du) - \int_0^t \int_{\Gamma} g\left(\frac{[2^n s]}{2^n}, u\right) \lambda(ds, du) \right| \\ & \leq 2\varepsilon t + 2\text{nd term}. \end{aligned}$$

Appealing to (2.15), we see that the 2nd term tends to 0, as $k \rightarrow \infty$, for any n . Hence we can conclude (2.16).

Let $\zeta_i, i = 1, 2$, be \tilde{M}_T (or $\hat{M}([0, T] \times \Gamma)$)-valued random variables, which may be defined on different probability spaces. ν_i denotes the probability distribution of ζ_i . So, ν_i is a probability on $(\tilde{M}_T, \tilde{d}_T)$ (or $(\hat{M}([0, T] \times \Gamma), \hat{d}_T)$) respectively. Let \tilde{m}_T (or \hat{m}_T) denote the totality of \tilde{M}_T (or $\hat{M}([0, T] \times \Gamma)$)-valued random variables. We endow the following Prohorov metric \tilde{D}_T (or \hat{D}_T) on \tilde{m}_T (or \hat{m}_T resp.),

$$(2.17) \quad \begin{aligned} \tilde{D}_T(\zeta_1, \zeta_2) &= L(\nu_1, \nu_2) \\ \hat{D}_T(\zeta_1, \zeta_2) &= L(\nu_1, \nu_2). \end{aligned}$$

Since \tilde{M}_T and $\hat{M}([0, T] \times \Gamma)$ are compact metric spaces, $(\tilde{m}_T, \tilde{D}_T)$ and (\hat{m}_T, \hat{D}_T) are also compact spaces.

For $\mathcal{A} = (\Omega, F, \hat{P}, \xi, B, Y, q)$, we sometimes denote ξ by $\xi_{\mathcal{A}}$ and so on, when any confusion might occur. Let $X(= X_{\mathcal{A}})$ be a solution of (2.1) for \mathcal{A} . Then (X, ξ, B, Y, q) becomes a $M_T = C([0, T] \rightarrow R^n) \times R^n \times C([0, T] \rightarrow R^n) \times ([0, T] \rightarrow R^1) \times \hat{M}([0, T] \times \Gamma)$ valued random variable. Endowing M_T

with a usual metric d_T (= sum of metric of each component) M_T turns out as a complete separable metric space. Let m_T denote the totality of M_T -valued random variables and D_T the Prohorov metric on m_T . Hereafter we denote (X, ξ, B, Y, q) by (X, \mathcal{A}) for simplicity if no confusion occurs. We also say that $\mathcal{A}_n \rightarrow \mathcal{A}$ (in Prohorov topology), if $(\xi_n, B_n, Y_n, q_n) \rightarrow (\xi, B, Y, q)$ in Prohorov topology. Ω_n, F_n, P_n can also depend on n . $\xi_{\mathcal{A}}, B_{\mathcal{A}}$ and $(Y_{\mathcal{A}}, q_{\mathcal{A}})$ are independent for any $\mathcal{A} \in \mathfrak{A}$ and $B_{\mathcal{A}}$ is a Brownian motion. Therefore we have

PROPOSITION 2.2. $\mathcal{A}_n \rightarrow \mathcal{A}$, iff $\xi_n \rightarrow \xi$ in law and $(Y_n, q_n) \rightarrow (Y, q)$ in Prohorov topology.

Now we put the set $\mathcal{P}_\mu =$ totality of probability distributions of $(Y_{\mathcal{A}}, q_{\mathcal{A}})$, $\mathcal{A} \in \mathfrak{A}(\mu)$, $\mathfrak{A}(\mu)$ defined later (3.8). Since $\xi_{\mathcal{A}}, B_{\mathcal{A}}$ and $(Y_{\mathcal{A}}, q_{\mathcal{A}})$ are independent for $\mathcal{A} \in \mathfrak{A}$, \mathcal{P}_μ does not depend on μ , say \mathcal{P} . Moreover $\pi \in \mathcal{P}$, iff π is a probability on $C([0, T] \rightarrow R^1) \times \hat{M}([0, T] \times \Gamma)$ such that the first component y is a Brownian motion under π and its increments $y(t) - y(s)$ is independent of $\sigma_s(y, \lambda)$ for $t > s$, where λ is the second component, (see § 2 of Fleming-Pardoux [8]). Since $C([0, T] \rightarrow R^1) \times \hat{M}([0, T] \times \Gamma)$ is a product of complete separable metric space, it becomes a complete separable metric space. So we introduce the Prohorov topology on \mathcal{P} . Then \mathcal{P} is a compact metric space, because the first component is a Brownian motion and $\hat{M}([0, T] \times \Gamma)$ is a compact metric space. Now we have

PROPOSITION 2.3.

i) $\mathcal{P} =$ totality of probability distribution of $(Y_{\mathcal{A}}, q_{\mathcal{A}})$, $\mathcal{A} \in \mathfrak{A}$, is a compact metric space with Prohorov metric.

ii) $\mathcal{P} =$ totality of probability distribution of $(Y_{\mathcal{A}}, q_{\mathcal{A}})$, $\mathcal{A} \in \mathfrak{A}(\mu)$, for any μ .

iii) For $\mathcal{A}, \mathcal{A}' \in \mathfrak{A}(\mu)$, $D_T((X_{\mathcal{A}}, \mathcal{A}), (X_{\mathcal{A}'}, \mathcal{A}')) = 0$ iff probability distribution of $(Y_{\mathcal{A}}, q_{\mathcal{A}}) =$ probability distribution of $(Y_{\mathcal{A}'}, q_{\mathcal{A}'})$.

§ 3. Existence of optimal relaxed control

Let N be a compact subset of probability measures on R^n with Prohorov metric. Put $P_\eta =$ Probability law of η and

$$\mathfrak{N} = \{(X, \mathcal{A}); P_{\xi_{\mathcal{A}}} \in N\}.$$

PROPOSITION 3.1. \mathfrak{N} is a compact subset of (m_T, D_T) .

Proof. By the condition (A1), $\{X_{\mathcal{A}}; P_{\xi_{\mathcal{A}}} \in N\}$ is totally bounded in Prohorov topology. $B_{\mathcal{A}}$ and $Y_{\mathcal{A}}$ are Brownian motions for any $\mathcal{A} \in \mathfrak{A}$. Since \hat{m}_T is compact, $\{q_{\mathcal{A}}; \mathcal{A} \in \mathfrak{A}\}$ is totality bounded. Therefore

$$\mathfrak{N} = \{(X, \mathcal{A}); P_{\xi_{\mathcal{A}}} \in N\}$$

is a totally bounded subset of (m_T, D_T) .

Now we will show that \mathfrak{N} is closed. Let (X_k, \mathcal{A}_k) , $k = 1, 2, \dots$ be a Cauchy sequence. Using Skorobod's theorem, we can construct $(X_k^*, \xi_k^*, B_k^*, Y_k^*, q_k^*)$ and $(X^*, \xi^*, B^*, Y^*, q^*)$ on a probability space $(\Omega^*, F^*, \dot{P}^*)$, so that

(3.1) $(X_k^*, \xi_k^*, B_k^*, Y_k^*, q_k^*)$ has the same probability law as $(X_k, \xi_k, B_k, Y_k, q_k)$, $k = 1, 2, \dots$.

(3.2) As $k \rightarrow \infty$, $(X_k^*, \xi_k^*, B_k^*, Y_k^*, q_k^*)$ converges to $(X^*, \xi^*, B^*, Y^*, q^*)$ in d_T metric, \dot{P}^* -almost surely.

Hence ξ^*, B^* and (Y^*, q^*) are independent and B^* and Y^* are Brownian motions. Moreover we see that, for a.a. $\omega(\dot{P}^*)$, $q^*(\cdot, \omega) \in \dot{M}([0, T] \times \Gamma)$ and $\hat{d}_T(q_k^*(\cdot, \omega), q^*(\cdot, \omega))$ tends to 0 as $k \rightarrow \infty$, by virtue of Proposition 2.2. On the other hand (2.15) implies that $q^*(t, A, \cdot)$ is F^* -measurable. Since $q(t, A, \omega)$ is continuous in t , $q^*(\cdot, A, \cdot)$ is $B_1[0, T] \times F^*$ -measurable. Namely q^* satisfies the conditions (v) and (vi). (3.2) again tells us that $Y^*(t) - Y^*(r_i)$ is independent of $\sigma_s(Y^*, q^*)$ whenever $s \leq r_i \leq t$. Since Y^* has continuous paths, this implies that $Y^*(t) - Y^*(s)$ is independent of $\sigma_s(Y^*, q^*)$. Therefore $\mathcal{A}^* = (\Omega^*, F^*, \dot{P}^*, \xi^*, B^*, Y^*, q^*) \in \mathfrak{A}$.

Next we will show that X^* is a solution of (2.1) for \mathcal{A}^* .

$$(3.3) \quad \left| \int_0^t \int_{\Gamma} \gamma(X_k^*(s, \omega), u) q_k^*(ds, du, \omega) - \int_0^t \int_{\Gamma} \gamma(X^*(s, \omega), u) q^*(ds, du, \omega) \right| \\ \leq \int_0^t \int_{\Gamma} |\gamma(X_k^*(s, \omega), u) - \gamma(X^*(s, \omega), u)| q_k^*(ds, du, \omega) \\ + \left| \int_0^t \int_{\Gamma} \gamma(X^*(s, \omega), u) (q_k^*(ds, du, \omega) - q^*(ds, du, \omega)) \right|$$

$$\text{the 1st term} \quad \leq K \int_0^t |X_k^*(s, \omega) - X^*(s, \omega)| q_k^*(ds, \Gamma, \omega) \\ = K \int_0^t |X_k^*(s, \omega) - X^*(s, \omega)| ds.$$

Since $X_k^*(\cdot, \omega)$ converges to $X^*(\cdot, \omega)$ uniformly in $[0, T]$, the 1st term

converges to 0, as $k \rightarrow \infty$. Appealing to Remark of Proposition 2.1, the 2nd term converges to 0, as $k \rightarrow \infty$. So, we have

$$(3.4) \quad \int_0^t \int_{\mathcal{F}} \gamma(X_k^*(s, \omega), u) q_k^*(ds, du, \omega) \longrightarrow \int_0^t \int_{\mathcal{F}} \gamma(X^*(s, \omega)) q^*(ds, du, \omega).$$

Using a routine method we get

$$(3.5) \quad \int_0^t \alpha(X_k^*(s)) dB_k^*(s) \longrightarrow \int_0^t \alpha(X^*(s)) dB^*(s) \quad \text{in proba } (\dot{P}^*).$$

From (3.4) and (3.5), we conclude that X^* is a solution of (2.1) for \mathcal{A}^* . This completes the proof of Proposition 3.1.

COROLLARY. *If $\mathcal{A}_k \rightarrow \mathcal{A}$, then $(X_k, \mathcal{A}_k) \rightarrow (X, \mathcal{A})$ in D_T .*

Let f and h be bounded and uniformly continuous functions on R^n . Define a pay-off function $J(\mathcal{A})$ as follows,

$$(3.6) \quad J(\mathcal{A}) = \dot{E}f(X_{\mathcal{A}}(T))L(T, \mathcal{A})$$

where \dot{E} stands for the expectation in (Ω, F, \dot{P}) , and

$$(3.7) \quad L(T, \mathcal{A}) = \exp \left(\int_0^T h(X(s)) dY(s) - \frac{1}{2} \int_0^T |h(X(s))|^2 ds \right)$$

where $X = X_{\mathcal{A}}$ and $Y = Y_{\mathcal{A}}$.

For a probability measure μ , we denote

$$(3.8) \quad \mathfrak{A}(\mu) = \{\mathcal{A} \in \mathfrak{A}; P_{\varepsilon_{\mathcal{A}}} = \mu\}$$

i.e. the set of all admissible system where initial distribution equals to μ .

For a given μ we want to maximize $J(\mathcal{A})$ by a suitable choice of $\mathcal{A} \in \mathfrak{A}(\mu)$.

THEOREM 2. *There exists an optimal admissible system $\tilde{\mathcal{A}} \in \mathfrak{A}(\mu)$, that is*

$$(3.9) \quad \sup_{\mathcal{A} \in \mathfrak{A}(\mu)} J(\mathcal{A}) = J(\tilde{\mathcal{A}}).$$

Proof. Let $\mathcal{A}_k \in \mathfrak{A}(\mu)$ be approximately optimal, i.e.

$$(3.10) \quad \lim_{k \rightarrow \infty} J(\mathcal{A}_k) = \sup_{\mathcal{A} \in \mathfrak{A}(\mu)} J(\mathcal{A}).$$

By virtue of Proposition 3.1, some subsequence $(X_{k_i}, \mathcal{A}_{k_i})$ converges to (X, \mathcal{A}) in Prohorov topology. For simplicity we may assume $(X_k, \mathcal{A}_k) \rightarrow$

(X, \mathcal{A}) as $k \rightarrow \infty$. Again Skorobod's theorem tells us that their suitable version satisfy (3.1) and (3.2). So we again assume that (X_k, \mathcal{A}_k) and (X, \mathcal{A}) satisfy (3.1) and (3.2), since $J(\mathcal{A})$ depends on only probability law.

From boundedness of f and h , we have

$$(3.11) \quad \mathring{E}(f(X_k(T))L(T, \mathcal{A}_k))^2 \leq \|f\|^2 e^{2T\|h\|}, \quad k = 1, 2, \dots$$

Hence $\{f(X_k(T))L(T, \mathcal{A}_k), k = 1, 2, \dots\}$ is uniformly integrable. On the other hand $L(T, \mathcal{A}_k)$ tends to $L(T, \mathcal{A})$ in proba. Appealing to the convergence theorem we get

$$(3.12) \quad \lim_{k \rightarrow \infty} J(\mathcal{A}_k) = J(\mathcal{A}).$$

Combining (3.12) with (3.10), we complete the proof.

Remark. Appealing to Corollary of Proposition 3.1, we see that if $\mathcal{A}_k \rightarrow \mathcal{A}$, then $J(\mathcal{A}_k) \rightarrow J(\mathcal{A})$.

Now we treat the following case; $r(x, u) = b_1(x) + b_2(x)u$ where $b_i(x)$ is $n \times k$ matrix.

THEOREM 3. *If $r(x, u) = b_1(x) + b_2(x)u$, then $q = q_{\mathcal{A}}$ can be replaced by a Γ -valued $\sigma_i(q)$ -progressively measurable process U (i.e. usual admissible control under partial observation). That is, $X = X_{\mathcal{A}}$ is a unique solution of the following S.D.E.*

$$(3.13) \quad \begin{cases} dX(t) = \alpha(X(t))dB(t) + r(X(t), U(t))dt \\ X(0) = \xi \end{cases}$$

where $B = B_{\mathcal{A}}$ and $\xi = \xi_{\mathcal{A}}$.

Proof. Our required U is obtained by the following lemma.

LEMMA. *There exists a $\sigma_i(q)$ -progressively measurable Γ -valued process U such that*

$$(3.14) \quad \int_0^t \int_{\Gamma} uq(ds, du) = \int_0^t U(s)ds, \quad \text{for } \forall t \leq T.$$

Proof. Define U_k as follows

$$(3.15) \quad U_k(t, \omega) = \begin{cases} 2^k \int_{\Gamma} u(q(t, du, \omega) - q(t - 2^{-k}, du, \omega)), & \text{for } t \geq 2^{-k} \\ \frac{1}{t} \int_{\Gamma} uq(t, du, \omega), & \text{for } 0 < t < 2^{-k}. \end{cases}$$

Since Γ is convex compact, $U_k(t, \omega) \in \Gamma$ and $\sigma_t(q)$ -progressively measurable. Moreover the compactness of Γ tells us that $\{U_k, k = 1, 2, \dots\}$ is weakly totally bounded in $L^2([0, T] \times \Omega)$. Hence some subsequence converges weakly and their suitable convex combinations converge strongly; say $\sum_{p=\ell}^{N\ell} \sigma_p^\ell U_{n_p}(t, \omega)$ converges to $U(t, \omega)$ in $L^2([0, T] \times \Omega)$, as $\ell \rightarrow \infty$. So U is a Γ -valued $\sigma_t(q)$ -progressively measurable process.

On the other hand the definition of U_k implies

$$(3.16) \quad \int_0^t U_k(s, \omega) ds \rightarrow \int_{\Gamma} uq(t, du, \omega) \left(= \int_0^t \int_{\Gamma} uq(ds, du, \omega) \right).$$

as $k \rightarrow \infty$. Taking the convex combination of U_{n_p} , we can conclude that U satisfies (3.14) by bounded convergence theorem.

Now we return to the proof of Theorem 3. Since ξ , B and U are independent, (3.13) has a unique solution. So it is enough to show that $X = X_{\mathcal{A}}$ satisfies (3.13). By the Lemma we can see

$$(3.17) \quad \int_0^t b_2(X(s)) \int_{\Gamma} uq(ds, du) = \int_0^t b_2(X(s)) U(s) ds.$$

Using " $\int_0^t b_1(X(s)) q(ds, du) = \int_0^t b_1(X(s)) ds$ ", we have

$$(3.18) \quad \int_0^t \int_{\Gamma} r(X(s), u) q(ds, du) = \int_0^t r(X(s), U(s)) ds.$$

This completes the proof of Theorem 3.

DEFINITION 3. A Γ -valued process U is called an admissible control under partial observation, if ξ , B and (Y, U) are independent and $Y(t) - Y(s)$ is independent of $\sigma_s(Y, U)$. Precisely speaking $\mathcal{A}_U = (\Omega, F, \dot{P}, \xi, B, Y, U)$ is called an admissible usual system.

An admissible control U can be regarded as the following relaxed control q ,

$$(3.19) \quad q(t, A, \omega) = \int_0^t \delta_{U(s, \omega)}(A) ds = |s \leq t; U(s, \omega) \in A|$$

where δ_a is the δ -measure at a . Appealing to Theorems 2 and 3, we can derive,

COROLLARY. *If $r(x, u) = b_1(x) + b_2(x)u$, then there exists an optimal admissible usual system $\tilde{\mathcal{A}}_U$. That is,*

$$(3.20) \quad J(\tilde{\mathcal{A}}_U) = \sup_{\mathcal{A}_U: \text{ad. usual sys}} J(\mathcal{A}_U) = \sup_{\mathcal{A} \in \mathfrak{A}(u)} J(\mathcal{A}).$$

This fact was directly proved by Hausmann [9] and in a slightly different form by Fleming-Pardoux [8].

§4. Approximation by usual controls

For $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, Y, q)$ we define P_n by

$$(4.1) \quad P_n(t, A, \omega) = \begin{cases} \frac{1}{t} q(t, A, \omega) & 0 < t < 2^{-n} \\ (q(t, A, \omega) - q(t - 2^{-n}, A, \omega))2^n, & 2^{-n} \leq t, \end{cases}$$

namely P_n is an approximate time derivative of q . $P_n(\cdot, A, \cdot)$ is $\sigma_t(q)$ -progressively measurable and $P_n(t, \cdot, \omega)$ is a probability on Γ . Define q_n by

$$(4.2) \quad q_n(t, A, \omega) = \int_0^t P_n(s, A, \omega) ds.$$

Then q_n satisfies the conditions (v) and (vi) and $\mathcal{A}_n = (\Omega, F, \dot{P}, \xi, B, Y, q_n) \in \mathfrak{U}$. Since we have

$$(4.3) \quad |q_n(t, A, \omega) - q(t, A, \omega)| \leq 2^{-n} + 2^n \int_{t-2^{-n}}^t |q(s, A, \omega) - q(t, A, \omega)| ds, \\ \text{for a.a. } \omega(\dot{P}),$$

the condition (iv) implies, as $n \rightarrow \infty$,

$$(4.4) \quad \sup_A |q_n(t, A, \omega) - q(t, A, \omega)| \rightarrow 0, \quad \text{uniformly on } [0, T].$$

and

$$(4.5) \quad \hat{d}_T(q_n(\cdot, \omega), q(\cdot, \omega)) \rightarrow 0.$$

Fix $u_0 \in \Gamma$ arbitrarily and define $P_{n,k}$ by

$$(4.6) \quad P_{n,k}(t, A, \omega) = \begin{cases} P_n\left(\frac{[2^k t]}{2^k}, A, \omega\right), & \text{for } t \geq 2^{-k} \\ \delta_{u_0}(A) & \text{for } t < 2^{-k}. \end{cases}$$

Then $P_{n,k}$ is a step function in the time variable t . Put $q_{n,k}$ as follows.

$$(4.7) \quad q_{n,k}(t, A, \omega) = \int_0^t P_{n,k}(s, A, \omega) ds.$$

We call $q_{n,k}$ a switching relaxed control with interval 2^{-k} . It is clear that $(\Omega, F, \dot{P}, \xi, B, Y, q_{n,k}) \in \mathfrak{U}$ and

$$(4.8) \quad |q_{n,k}(t, A, \omega) - q_n(t, A, \omega)| \leq 2^{n-k-1}t + 2^{-k}.$$

Therefore we get

$$(4.9) \quad \limsup_{k \rightarrow \infty} \sup_{t \leq T, A} |q_{n,k}(t, A, \omega) - q_n(t, A, \omega)| = 0.$$

Now we conclude the following proposition,

PROPOSITION 4.1. *For $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, Y, q) \in \mathfrak{X}$, there exists an approximate sequence of switching relaxed control q_k with interval 2^{-k} , such that $q_k(\cdot, A, \cdot)$ is $\sigma_t(q)$ -progressively measurable and moreover*

$$(4.10) \quad \limsup_{k \rightarrow \infty} \sup_{0 \leq t \leq T, A} |q_k(t, A) - q(t, A)| = 0, \quad \dot{P}\text{-almost surely}$$

and

$$(4.11) \quad \lim_{k \rightarrow \infty} \hat{d}_T(q_k, q) = 0, \quad \dot{P}\text{-almost surely.}$$

Putting $\mathcal{A}_k = (\Omega, F, \dot{P}, \xi, B, Y, q_k)$, we can see the following corollary, by virtue of the Remark of Theorem 2.

COROLLARY. *There exists an approximate admissible switching system \mathcal{A}_k , such that*

$$(4.12) \quad \mathcal{A}_k \longrightarrow \mathcal{A}, \quad \text{as } k \rightarrow \infty.$$

Hence $J(\mathcal{A}_k)$ converges to $J(\mathcal{A})$.

THEOREM 4. *There exists a Γ -valued $\sigma_t(q)$ -progressively measurable process U_k , such that*

$$(4.13) \quad q_k(t, A, \omega) = \int_0^t \delta_{U_k(s, \omega)}(A) ds$$

approximates q in the following sense; $\mathcal{A}_k = (\Omega, F, \dot{P}, \xi, B, Y, q_k)$ satisfies (4.12).

Proof. By the Corollary of Proposition 4.1, we may assume that \mathcal{A} is an admissible switching system with interval 2^{-N} . Appealing to a Chattering Lemma [5], we will construct our desired U_{k_i} in the following way.

Let $\{u_1, \dots, u_m\}$ be an ε -net of Γ , and $V_1, \dots, V_m \in \mathcal{B}(\Gamma)$ a partition of Γ such that

$$(4.14) \quad |u_i - u| < \varepsilon \quad \text{for } \forall u \in V_i.$$

Since a given q is a switching relaxed control, it can be written by

$$q(t, A, \omega) = \int_0^t p(s, A, \omega) ds$$

with p of step function in s . Define $\hat{p} = \hat{p}_\varepsilon$ and $\hat{q} = \hat{q}_\varepsilon$ as follows,

$$(4.15) \quad \hat{p}(s, \{u_i\}, \omega) = p(s, V_i, \omega)$$

and

$$(4.16) \quad \hat{q}(t, \{u_i\}, \omega) = \int_0^t \hat{p}(s, \{u_i\}, \omega) ds.$$

Then $\hat{p}(t, \cdot, \omega)$ is a discrete probability on Γ and for $\forall g \in C(\Gamma)$

$$(4.17) \quad \int_{\Gamma} g(u) \hat{p}_\varepsilon(t, du, \omega) \longrightarrow \int_{\Gamma} g(u) p(t, du, \omega), \quad \text{as } \varepsilon \downarrow 0.$$

Define $\theta_i, i = 0, \dots, m$ as follows: Let $j^{2^{-N}} \leq s_1 < s_2 < (j+1)2^{-N}$,

$$(4.18) \quad \begin{aligned} \theta_0(\omega) &= s_1 \\ \theta_i(\omega) &= \sum_{\ell=1}^i \int_{s_1}^{s_2} \hat{p}(t, \{u_i\}, \omega) dt + s_1, \quad i = 1, \dots, m. \end{aligned}$$

Then $s_1 = \theta_0(\omega) \leq \theta_1(\omega) \leq \dots \leq \theta_m(\omega) = s_2$ and

$$(4.19) \quad \begin{aligned} \int_{s_1}^{s_2} \int_{\Gamma} g(u) \hat{q}(dt, du, \omega) &= \int_{s_1}^{s_2} \int_{\Gamma} g(u) \hat{p}(t, du, \omega) dt \\ &= \sum_{i=1}^m g(u_i) \int_{s_1}^{s_2} \hat{p}(t, \{u_i\}, \omega) dt = \sum_{i=1}^m g(u_i) (\theta_i(\omega) - \theta_{i-1}(\omega)) \\ &= \int_{s_1}^{s_2} g(U(t, \omega)) dt = \int_{s_1}^{s_2} \int_{\Gamma} g(u) \delta_{U(t, \omega)}(du) dt \end{aligned}$$

where

$$(4.20) \quad U(t, \omega) = U_{\varepsilon, s_1, s_2}(t, \omega) = u_i \quad \text{on } [(\theta_{i-1}(\omega), \theta_i(\omega))].$$

Therefore $U_{\varepsilon, s_1, s_2}(t)$ is $\sigma_{j^{2^{-N}}(q)}$ -measurable. Putting $\varepsilon = 2^{-k}, s_1 = \ell 2^{-k} (k > N)$, we define U_k by

$$(4.21) \quad \begin{aligned} U_k(t, \omega) &= U_{2^{-k}, \ell 2^{-k}, (\ell+1)2^{-k}}(t, \omega) \\ &\text{for } \ell 2^{-k} \leq t < (\ell+1)2^{-k}, \ell = 0, 1, 2, \dots \end{aligned}$$

Consider the SDE

$$(4.22) \quad \begin{cases} d\xi_k(t) = \alpha(\xi_k(t)) dB(t) + \gamma(\xi_k(t), U_k(t)) dt \\ \xi_k(0) = \xi. \end{cases}$$

If we regard $U_k(t)$ as $\delta_{U_k(t)}$, then ξ_k turns out a solution of (2.1) for $\mathcal{A}_k = (\Omega, F, \hat{P}, \xi, B, Y, q_k)$ where $q_k(t, A, \omega) = \int_0^t \delta_{U_k(s, \omega)}(A) ds$. Moreover (4.19) means

$$(4.23) \quad q_k(\ell 2^{-k}, \cdot, \omega) = \hat{q}_{2^{-k}}(\ell 2^{-k}, \cdot, \omega), \quad \ell = 0, 1, \dots, [2^k T].$$

Hence, combining with (4.17), we can see that, as $k \rightarrow \infty$,

$$(4.24) \quad \hat{d}_T(q_k, q) \longrightarrow 0, \quad \hat{P}\text{-almost surely.}$$

Evidently this completes the proof.

§ 5. Continuity of conditional expectation

According to [8] we define L pathwise. Hereafter we assume the following smoothness on h .

$$(A3) \quad h, \frac{\partial h}{\partial x_i}, \frac{\partial^2 h}{\partial x_i \partial x_j} \quad i, j = 1, \dots, n \text{ are bounded and uniformly continuous.}$$

Putting $X = X_{\omega}$, Ito's formula tells us that

$$(5.1) \quad \begin{aligned} \int_0^T h(X(t)) dY(t) &= h(X(T))Y(T) - \int_0^T Y(t) dh(X(t)) \\ &= h(X(T))Y(T) - \sum_{ij} \int_0^T Y(t) \frac{\partial h}{\partial x_i}(X(t)) \alpha_{ij}(X(t)) dB_j \\ &\quad - \int_0^T Y(t) A(t, q) h(X(t)) dt \end{aligned}$$

where

$$(5.2) \quad A(t, q)h = \frac{1}{2} \sum a_{ij}(x) \frac{\partial^2 h}{\partial x_i \partial x_j} + \sum R_i(t, x, \omega; q) \frac{\partial h}{\partial x_i}$$

with

$$(5.3) \quad R(t, x, \omega; q) = \overline{\lim}_{n \rightarrow \infty} 2^n \int_r \gamma(x, u)(q(t, du, \omega) - q(t - 2^{-n}, du, \omega)).$$

So, R is $\sigma_t(q)$ -progressively measurable for any x , and Lipschitz continuous with respect to x . Moreover for any (x, ω) ,

$$(5.4) \quad R(t, x, \omega; q) = \frac{\partial}{\partial t} \int_r \gamma(x, u) q(t, du, \omega) \quad \text{for a.a. } t.$$

For $y \in C([0, T] \rightarrow R^n)$ and $\lambda \in \hat{M}([0, T] \times \Gamma)$ we define \mathcal{L} by

$$(5.5) \quad \begin{aligned} &\mathcal{L}(\theta, \xi, B, y, \lambda) \\ &= \exp \left[y(\theta)h(\gamma(\theta)) - \int_0^\theta y(s)A(s, \lambda)h(\gamma(s))ds - \frac{1}{2} \int_0^\theta |h(\gamma(s))|^2 ds \right. \\ &\quad \left. - \sum_{ij} \int_0^s y(s) \frac{\partial h}{\partial x_i}(\gamma(s)) \alpha_{ij}(\gamma(s)) dB_j(s) \right] \end{aligned}$$

where η is a solution of S.D.E.

$$(5.6) \quad \begin{cases} d\eta(t) = \alpha(\eta(t))dB(t) + \int_{\Gamma} \gamma(y(t), u)\lambda(dt, du) \\ \eta(0) = \xi. \end{cases}$$

Applying a successive approximation, we can see that η is a Borel function of ξ, B and λ . Hence \mathcal{L} is Borel measurable with respect to $\theta, \xi, B, y, \lambda$. From (A1) and (A3) we have the following evaluation,

$$(5.7) \quad \begin{aligned} & e^{-F(y, \theta)} \exp \left[- \sum_{ij} \int_0^{\theta} y(t) \frac{\partial h}{\partial x_i}(\eta(t)) \alpha_{ij}(\eta(t)) dB_j(t) \right. \\ & \quad \left. - \frac{1}{2} \int_0^t \sum_j \left(y(t) \sum_i \frac{\partial h}{\partial x_i}(\eta(t)) \alpha_{ij}(\eta(t)) \right)^2 dt \right] \\ & \leq \mathcal{L}(\theta, \xi, B, y, \lambda) \leq e^{F(y, \theta)} \exp \left[- \sum_{ij} \int_0^{\theta} \dots dB_j(t) - \frac{1}{2} \int_0^t \dots dt \right] \end{aligned}$$

where

$$F(y, \theta) = K_1(\sup_{t \leq \theta} |y(t)| + 1)^2(\theta + 1).$$

Since $(C[0, T] \rightarrow R^n, \|\cdot\|)$ and $(\hat{M}[0, T] \times \Gamma, \hat{d}_T)$ are complete separable metric spaces, the regular conditional probability $\hat{P}((X, \xi, B, Y, q) \in \cdot / Y = y, q = \lambda)$ exists. This regular conditional probability is nothing but the probability distribution of $(\eta, \xi, B, y, \lambda)$, because (ξ, B) and (Y, q) are independent. Putting $\mu = P_{\xi}$ (= probability distribution of ξ), we have a version of conditional expectation as follows.

$$(5.8) \quad \begin{aligned} \hat{E}(f(X(\theta))L(\theta, \mathcal{A}) / Y = y, q = \lambda) &= \hat{E}f(\eta(\theta))\mathcal{L}(\theta, \xi, B, y, \lambda) \\ &= \int \hat{E}f(\eta(\theta, x))\mathcal{L}(\theta, x, B, y, \lambda) d\mu(x), \quad \text{for bounded Borel } f, \end{aligned}$$

where $\eta(\theta, x)$ is a solution of (5.6) with $\eta(0, x) = x$. The right side of (5.8) is Borel measurable with respect to θ, y, λ , which depends on f and μ . So we denote the right side of (5.8) by $C(\theta, y, \lambda, \mu, f)$. Moreover $C(\theta, y, \lambda, \mu, f)$ depends on the value of y and λ up to time θ . Stressing μ we denote \hat{E} by \hat{E}_{μ} . That is,

$$(5.9) \quad \begin{aligned} C(\theta, Y, q, \mu, f) &= \hat{E}_{\mu}(f(X(\theta))L(\theta, \mathcal{A}) / \sigma_T(Y, q)) \\ &= \hat{E}_{\mu}(f(X(\theta))L(\theta, \mathcal{A}) / \sigma_{\theta}(Y, q)), \quad \hat{P}\text{-almost surely.} \end{aligned}$$

Using (5.7) we have

$$(5.10) \quad e^{-F(Y, \theta)} \leq C(\theta, Y, q, \mu, 1) \leq e^{F(Y, \theta)}.$$

Now we define $C(\theta, Y, q, \nu, f)$ for a positive measure ν as follows

$$(5.11) \quad C(\theta, Y, q, \nu, f) = \|\nu\| C\left(\theta, Y, q, \frac{\nu}{\|\nu\|}, f\right)$$

where $\|\nu\| = \nu(R^n)$ and we apply the same notations \dot{E}_ν for a general positive measure ν , that is

$$(5.12) \quad \dot{E}_\nu(f(X(\theta))L(\theta, \mathcal{A})/\sigma_\theta(Y, q)) = C(\theta, Y, q, \nu, f)$$

with $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, Y, q)$ for $P_\xi = \nu/\|\nu\|$. The left side of (5.12) stands for

$$(5.13) \quad \dot{E}_\nu(f(X(\theta))L(\theta, \mathcal{A})/\sigma_\theta(Y, q)) = \|\nu\| \dot{E}_{\nu/\|\nu\|}(f(X(\theta))L(\theta, \mathcal{A})/\sigma_\theta(Y, q)).$$

Since ξ is independent of (B, Y, q) , the right side of (5.13) does not depend on a special choice of ξ .

Define $A(\theta, y, \lambda, \nu)(A)$ by

$$(5.14) \quad A(\theta, y, \lambda, \nu)(A) = C(\theta, y, \lambda, \nu, \chi_A), \quad A \in \mathcal{B}_n.$$

Then $A(\theta, y, \lambda, \nu)$ is a positive measure on R^n and for any bounded Borel function f .

$$(5.15) \quad \langle f, A(\theta, y, \lambda, \nu) \rangle = C(\theta, y, \lambda, \nu, f)$$

where $\langle f, A \rangle = \int_{R^n} f(x)A(dx)$. From (5.10) we see

$$(5.16) \quad \|\nu\| e^{-F(y, \theta)} \leq \|A(\theta, y, \lambda, \nu)\| \leq \|\nu\| e^{F(y, \theta)}.$$

On the other hand $\mathcal{L}(\theta, \xi, B, y, \lambda)$ is continuous in θ , \dot{P} -almost surely, and (5.7) implies the uniform integrability of $\{\mathcal{L}(\theta, \xi, B, y, \lambda), \theta \in [0, T]\}$. Hence $\|A(\theta, y, \lambda, \nu)\| = \|\nu\| \dot{E}\mathcal{L}(\theta, \xi, B, y, \lambda)$ is continuous in θ .

Define a metric Δ on $M(R^n)$ (= totality of positive measure on R^n) as follows

$$(5.17) \quad A(\mu, \nu) = L\left(\frac{\mu}{\|\mu\|}, \frac{\nu}{\|\nu\|}\right) + \left| \|\mu\| - \|\nu\| \right| + \left| \frac{1}{\|\mu\|} - \frac{1}{\|\nu\|} \right|.$$

Then $(M(R^n), \Delta)$ is a complete separable metric space and $\nu_k, k = 1, 2, \dots$ is a Cauchy sequence, iff $\langle f, \nu_k \rangle$ converges for any $f \in C_b(R^n)$, as $k \rightarrow \infty$. and $\lim_{k \rightarrow \infty} \langle 1, \nu_k \rangle (= \|\nu_k\|) > 0$. Recalling Prohorov's theorem we have

PROPOSITION 5.1. *$N \supset M(R^n)$ is Δ -totally bounded, iff there exist positive constants c and c' and for $\forall \varepsilon > 0$ there is a compact subset $K_\varepsilon \subset R^n$ such that*

$$(5.18) \quad c' \leq \|\nu\| \leq c \quad \text{and} \quad \nu(K_\varepsilon^c) < \varepsilon \quad \text{for } \nu \in N.$$

Put $m =$ totality of $M(R^n)$ -valued random variables, which may be defined on different probability spaces. We endow the Prohorov metric on m , (called δ metric), namely

$$\delta(\zeta_1, \zeta_2) = L(\mu_1, \mu_2)$$

where μ_i is the probability distribution of ζ_i . Then (m, δ) is a complete separable metric space, because $(M(R^n), \Delta)$ is a complete separable metric space.

Concerning the continuity of A , we can prove the following theorem.

THEOREM 5. *If $y_k \rightarrow y$ in $C([0, T] \rightarrow R^1)$, $\nu_k \rightarrow \nu$ in Δ , $\lambda_k \rightarrow \lambda$ in \hat{d}_T and $\theta_k \rightarrow \theta$, then*

$$(5.19) \quad A(\theta_k, y_k, \lambda_k, \nu_k) \longrightarrow A(\theta, y, \lambda, \nu) \quad \text{in } \Delta.$$

Proof. Firstly we remark that

$$(5.20) \quad \int_0^s R(t, x, \lambda_k) dt \longrightarrow \int_0^s R(t, x, \lambda) dt.$$

Recalling the definition of R for $\lambda \in \hat{M}([0, T] \times \Gamma)$ (see (5.3)), we get, for any x ,

$$(5.21) \quad R(t, x, \lambda_k) = \lim_{\lambda \rightarrow \infty} \int_{\Gamma} \gamma(x, u) (\lambda_k(t, du) - \lambda_k(t - 2^{-n}, du)) 2^n$$

for a.a. t .

Hence the bounded convergence theorem implies

$$(5.22) \quad \begin{aligned} \int_0^s R(t, x, \lambda_k) dt &= \lim_{n \rightarrow \infty} \left[\int_0^s \int_{\Gamma} 2^n \gamma(x, u) \lambda_k(t, du) dt \right. \\ &\quad \left. - \int_{2^{-n}}^s \int_{\Gamma} 2^n \gamma(x, u) \lambda_k(t - 2^{-n}, du) dt \right] \\ &= \int_{\Gamma} \gamma(x, u) \lambda_k(s, du). \end{aligned}$$

Since $\lambda_k \rightarrow \lambda$ in \hat{d}_T , (5.22) means (5.20).

Consequently we can easily see

LEMMA. *If $\phi_k \rightarrow \phi$ and $\psi_k \rightarrow \psi$ in $C([0, T] \rightarrow R^n)$ and $C([0, T] \rightarrow R^1)$ respectively, then*

$$(5.23) \quad \int_0^s R(t, \phi_k(t), \lambda_k) \psi_k(t) dt \longrightarrow \int_0^s R(t, \phi(t), \lambda) \psi(t) dt.$$

Putting $A_k = A(\theta_k, y, \lambda_k, \nu_k)$ and $A = A(\theta, y, \lambda, \nu)$, we will show

$$(5.24) \quad \langle f, A_k \rangle \longrightarrow \langle f, A \rangle \quad \text{for } f \in C_b(R^n).$$

Consider the SDE on $(\Omega_k, F_k, \dot{P}_k)$

$$(5.25) \quad \begin{cases} d\eta_k(t) = \alpha(\eta_k(t))dB_k(t) + \int_{\Gamma} \gamma(\eta_k(t), u)\lambda_k(dt, du) \\ \eta_k(0) = \xi_k \end{cases}$$

and on (Ω, F, \dot{P})

$$(5.26) \quad \begin{cases} d\eta(t) = \alpha(\eta(t))dB(t) + \int_{\Gamma} \gamma(\eta(t), u)\lambda(dt, du) \\ \eta(0) = \xi \end{cases}$$

where ξ_k and ξ have probability distributions $\nu_k/\|\nu_k\|$ and $\nu/\|\nu\|$ respectively. Since $\{(\eta_k, \xi_k, B_k), k = 1, 2, \dots\}$ is totally bounded in Prohorov topology and any convergent subsequence tends to (η, ξ, B) in Prohorov topology, $(\eta_k, \xi_k, B_k), k = 1, 2, \dots$ itself converges to (η, ξ, B) in Prohorov topology. Appealing to Skorobod's theorem, we will assume that $\Omega_k = \Omega, F_k = F, \dot{P}_k = \dot{P}$ and \dot{P} -almost surely $\eta_k \rightarrow \eta$ in $C([0, T] \rightarrow R^n)$, $B_k \rightarrow B$ in $C([0, T] \rightarrow R^n)$ and $\xi_k \rightarrow \xi$ in R^n . Therefore the lemma guarantees

$$(5.27) \quad \int_0^{\theta_k} y_k(s)A(s, \lambda_k)h(\eta_k(s))ds \longrightarrow \int_0^{\theta} y(s)A(s, \lambda)h(\eta(s))ds, \\ \dot{P}\text{-almost surely.}$$

Furthermore, using a routine method we have

$$(5.28) \quad \int_0^{\theta_k} y_k(s) \frac{\partial h}{\partial x_i}(\eta_k(s))\alpha_{ij}(\eta_k(s))dB_{k,j}(s) \\ \longrightarrow \int_0^{\theta} y(s) \frac{\partial h}{\partial x_i}(\eta(s))\alpha_{ij}(\eta(s))dB_j(s) \quad \text{in proba } \dot{P}.$$

Hence we have

$$(5.29) \quad \mathcal{L}(\theta_k, \xi_k, B_k, y_k, \lambda_k) \longrightarrow \mathcal{L}(\theta, \xi, B, y, \lambda) \quad \text{in proba } \dot{P}.$$

Since (5.7) means the uniform integrability of $\{f(\eta_k(\theta_k))\mathcal{L}(\theta_k, \xi_k, B_k, y_k, \lambda_k), k = 1, 2, \dots\}$, (5.29) implies (5.24).

By virtue of Proposition 5.1, (5.16) and (5.20) guarantee the totally boundedness of $\{A_k, k = 1, 2, \dots\}$. Consequently, again (5.24) tells us that A_k converges to A in metric \mathcal{A} . This completes the proof of Theorem 5.

Now we apply this theorem to admissible systems.

THEOREM 6. Let $\mathcal{A}_k = (\Omega_k, F_k, \dot{P}_k, \xi_k, B_k, Y_k, q_k)$ and $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, Y, q)$ where ξ_k and ξ have probability distributions $\nu_k/\|\nu_k\|$ and $\nu/\|\nu\|$ respectively. If $(Y_k, q_k) \rightarrow (Y, q)$ in Prohorov topology, $\nu_k \rightarrow \nu$ in Δ and $\theta_k \rightarrow \theta$, then

$$(5.30) \quad \Lambda(\theta_k, Y_k, q_k, \nu_k) \longrightarrow \Lambda(\theta, Y, q, \nu) \quad \text{in metric } \delta.$$

Proof. By the assumption $\mathcal{A}_k \rightarrow \mathcal{A}$ in Prohorov topology. Hence $(X_k, \mathcal{A}_k) \rightarrow (X, \mathcal{A})$ in Prohorov metric D_T , by virtue of Corollary of Proposition 3.1. By Skorobod's theorem we can construct copies (X_k^*, \mathcal{A}_k^*) and (X^*, \mathcal{A}^*) of (X_k, \mathcal{A}_k) and (X, \mathcal{A}) respectively, so that $\Omega_k^* = \Omega^*$, $F_k^* = F^*$, $\dot{P}_k^* = \dot{P}^*$ and \dot{P}^* -almost surely $(X_k^*, \xi_k^*, B_k^*, Y_k^*, q_k^*) \rightarrow (X^*, \xi^*, B^*, Y^*, q^*)$ in d_T . For non-exceptional $\omega \in \Omega^*$, we put $\lambda_k = q_k^*(\cdot, \omega)$, $y_k = Y_k^*(\cdot, \omega)$, $\lambda = q^*(\cdot, \omega)$ and $y = Y^*(\cdot, \omega)$. Then y_k, λ_k, y and λ satisfy the condition of Theorem 5, Therefore we have

$$(5.31) \quad \Lambda(\theta_k, Y_k^*, q_k^*, \nu_k) \longrightarrow \Lambda(\theta, Y^*, q^*, \nu), \quad \dot{P}^*\text{-almost surely.}$$

On the other hand Theorem 5 tells us that the mapping $\Lambda(\theta, \cdot, \cdot, \nu); C([0, T] \rightarrow R^1) \times (\hat{M}([0, T] \times \Gamma)) \rightarrow M(R^n)$ is continuous. So $\Lambda(\theta, Y, q, \nu)$ is a random variable, i.e. $\Lambda(\theta, Y, q, \nu) \in m$. Consequently (5.31) implies (5.30).

Recalling Corollary of Theorem 4, we get

COROLLARY. For any $\mathcal{A} \in \mathfrak{A}(\mu)$, there exists an approximate admissible switching system $\mathcal{A}_k \in \mathfrak{A}(\mu)$, such that

$$(5.32) \quad \Lambda(\theta_k, Y_k, q_k, \nu) \longrightarrow \Lambda(\theta, Y, q, \nu) \quad \text{in metric } \delta$$

where $\mu = \nu/\|\nu\|$.

§ 6. Semigroup

Let C be the Banach lattice of the totality of bounded continuous mappings from $(M(R^n), \Delta)$ into R^1 , with supremum norm and the order \leq , i.e.

$$(6.1) \quad \phi \leq \psi \iff \phi(\nu) \leq \psi(\nu) \quad \text{for } \forall \nu \in M(R^n).$$

For $\nu \in M(R^n)$, $\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)$ and $\phi \in C$ we define J by

$$(6.2) \quad J(t, \mathcal{A}, \nu, \phi) = \dot{E}(\phi(\Lambda(t, Y, q, \nu)))$$

$$(6.3) \quad \begin{aligned} & \dot{E}[\dot{E}(\phi(\Lambda(t, Y, q, \nu)) | \sigma(Y, q))] \\ &= \int_{C([0, T] \rightarrow R^1) \times \hat{M}([0, T] \times \Gamma)} \dot{E}(\phi(\Lambda(t, y, \lambda, \nu))) \pi(dy, d\lambda) \end{aligned}$$

where π is the probability distribution of (Y, q) . Since $\dot{E}(\phi(\Lambda(t, y, \lambda, \nu)))$ depends only on ϕ, ν, t, y and $\lambda, J(t, \mathcal{A}, \nu, \phi)$ can be denoted by $J(t, \pi, \nu, \phi)$.

Define $S(t)\phi$ by

$$(6.4) \quad S(t)\phi(\nu) = \sup_{\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)} J(t, \mathcal{A}, \nu, \phi).$$

Then by Proposition 2.3 (ii) and (6.3) we have

$$(6.5) \quad S(t)\phi(\nu) = \sup_{\pi \in \mathcal{P}} J(t, \pi, \nu, \phi).$$

PROPOSITION 6.1. $J(t, \pi, \nu, \phi)$ is continuous in $(t, \pi, \nu) \in [0, T] \times \mathcal{P} \times M(R^n)$.

Proof. Let $t_k \rightarrow t, \pi_k \rightarrow \pi$ and $\nu_k \rightarrow \nu$ in their topologies. Take $\mathcal{A}_k \in \mathfrak{A}(\nu_k/\|\nu_k\|)$ (and $\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)$) such that the probability distribution of (y_k, q_k) (and (Y, q)) is π_k (and π respectively). Then $\mathcal{A}_k \rightarrow \mathcal{A}$ by Proposition 2.2. Therefore Theorem 6 guarantees that $\Lambda(t_k, \pi_k, \nu_k, \phi) \rightarrow \Lambda(t, \pi, \nu, \phi)$ in metric δ that is in the Prohorov topology. By Skorobod's theorem we can take a copy Λ_k^* of $\Lambda(t_k, \pi_k, \nu_k, \phi)$ and Λ^* of $\Lambda(t, \pi, \nu, \phi)$ so that Λ_k^* converges to Λ^* almost surely on $(\Omega^*, F^*, \dot{P}^*)$. Since ϕ is bounded continuous, we see that

$$(6.6) \quad \begin{cases} J(t_k, \pi_k, \nu_k, \phi) = \dot{E}_k \phi(\Lambda(t_k, Y_k, q_k, \nu_k)) = \dot{E}^* \phi(\Lambda_k^*) \\ J(t, \pi, \nu, \phi) = \dot{E} \phi(\Lambda(t, Y, q, \nu)) = \dot{E}^* \phi(\Lambda^*) \\ \dot{E}^* \phi(\Lambda_k^*) \rightarrow \dot{E}^* \phi(\Lambda^*) . \end{cases}$$

This completes the proof of Proposition 6.1.

Since \mathcal{P} is a compact metric space by Proposition 2.3 (i), we can conclude the following proposition.

PROPOSITION 6.2. $S(t)\phi \in C$ whenever $\phi \in C$. That is, $S(t)$ is a mapping from C into C . Recalling Corollary of Theorem 6, we see

$$(6.7) \quad S(t)\phi(\nu) = \sup_{\substack{\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|) \\ \mathcal{A}: \text{switching syst.}}} J(t, \mathcal{A}, \nu, \phi).$$

THEOREM 7. $S(t + \theta) = S(t)S(\theta), S(0) = \text{identity}$.

Proof. Consider the SDE on (Ω, F, \dot{P}) , for $\lambda \in \hat{M}([0, T] \times \Gamma)$

$$(6.8) \quad \begin{cases} d\gamma(t) = \alpha(\gamma(t))dB(t) + \int_{\Gamma} \gamma(\gamma(t), u)\lambda(dt, du) \\ \gamma(0) = \xi . \end{cases}$$

Since a solution $\eta(\cdot, \xi, B, \lambda)$ is unique, η satisfies the following relation

$$(6.9) \quad \eta(t + \theta, \xi, B, \lambda) = \eta(\theta, \eta(t, \xi, B, \lambda), B_t^+, \lambda_t^+)$$

where $B_t^+(s) = B(t + s) - B(t)$, $\lambda_t^+(s, A) = \lambda(t + s, A) - \lambda(t, A)$.

Using Ito's formula we get

$$(6.10) \quad \mathcal{L}(t + s, \xi, B, y, \lambda) = \mathcal{L}(t, \xi, B, y, \lambda) \mathcal{L}(s, \eta(t, \xi, B, \lambda), B_t^+, y_t^+, \lambda_t^+).$$

Define $v: [0, T] \times R^n \times C([0, T] \rightarrow R^1) \times \hat{M}([0, T] \times \Gamma) \times C_b(R^n) \rightarrow R^1$ by

$$(6.11) \quad v(t, x, y, \lambda, f) = \mathring{E}f(\eta(t, x, B, \lambda)) \mathcal{L}(t, x, B, y, \lambda)$$

where \mathring{E} of the right side stands for the expectation with respect to B , since the starting point x is not random. From (6.10) and (6.11) we have

$$(6.12) \quad \begin{aligned} v(t + s, x, y, \lambda, f) &= \mathring{E}f(\eta(t + s, x, B, \lambda)) \mathcal{L}(t + s, x, B, y, \lambda) \\ &= \mathring{E}[\mathcal{L}(t, x, B, y, \lambda) \mathring{E}(f(\eta(s, \eta(t, x, B, \lambda), B_t^+, \lambda_t^+) \\ &\quad \times \mathcal{L}(s, \eta(t, x, B, \lambda), B_t^+, y_t^+, \lambda_t^+ / \sigma_t(B)))]. \end{aligned}$$

Since $\eta(t, x, B, \lambda)$ is $\sigma_t(B)$ -measurable, we see

$$(6.13) \quad \begin{aligned} \mathring{E}(f(\eta(s, \eta(t, x, B, \lambda), B_t^+, \lambda_t^+) \mathcal{L}(s, \eta(t, x, B, \lambda), B_t^+, y_t^+, \lambda_t^+) / \sigma_t(B))) \\ = v(s, \eta(t, x, B, \lambda), y_t^+, \lambda_t^+, f) \end{aligned}$$

and, combining with (6.12) we get

$$(6.14) \quad \begin{aligned} v(t + s, x, y, \lambda, f) &= \mathring{E}v(s, \eta(t, x, B, \lambda), y_t^+, \lambda_t^+, f) \mathcal{L}(t, x, B, y, \lambda) \\ &= v(t, x, y, \lambda, v(s, \cdot, y_t^+, \lambda_t^+, f)). \end{aligned}$$

Recalling (5.8) and (5.15) we get

$$(6.15) \quad \langle f, A(t, y, \lambda, \nu) \rangle = \langle v(t, \cdot, y, \lambda), \nu \rangle, \quad f \in C_b(R^n).$$

Hence, by (6.14), we have

$$(6.16) \quad \begin{aligned} \langle f, A(t + s, y, \lambda, \nu) \rangle &= \langle v(t + s, \cdot, y, \lambda), \nu \rangle \\ &= \langle v(t, \cdot, y, \lambda, v(s, \cdot, y_t^+, \lambda_t^+, f)), \nu \rangle \\ &= \langle v(s, \cdot, y_t^+, \lambda_t^+, f), A(t, y, \lambda, \nu) \rangle \\ &= \langle f, A(s, y_t^+, \lambda_t^+, A(t, y, \lambda, \nu)) \rangle, \quad f \in C_b(R^n). \end{aligned}$$

Consequently

$$(6.17) \quad A(t + s, y, \lambda, \nu) = A(s, y_t^+, \lambda_t^+, A(t, y, \lambda, \nu)).$$

Since (6.17) holds for any $y \in C([0, T] \rightarrow R^1)$ and $\lambda \in \hat{M}([0, T] \times \Gamma)$, we have, for any $\mathcal{A} = (\Omega, F, \mathring{P}, \xi, B, Y, q) \in \mathfrak{A}(\nu / \|\nu\|)$,

$$(6.18) \quad \Lambda(t + s, Y, q, \nu) = \Lambda(s, Y_t^+, q_t^+, \Lambda(t, Y, q, \nu)), \dot{P}\text{-almost surely.}$$

This implies

$$(6.19) \quad \begin{aligned} J(t + s, \mathcal{A}, \nu, \phi) &= \dot{E}\phi(\Lambda(t + s, Y, q, \nu)) \\ &= \dot{E}[\dot{E}(\phi(\Lambda(t + s, Y, q, \nu))/\sigma_T(Y, q))] \\ &= \dot{E}\phi(\Lambda(s, Y_t^+, q_t^+, \Lambda(t, Y, q, \nu))) \\ &= \dot{E}\dot{E}(\phi(\Lambda(s, Y_t^+, q_t^+, \Lambda(t, Y, q, \nu))/\sigma_t(Y, q))). \end{aligned}$$

Under the regular conditional probability $\dot{P}(\cdot/\sigma_t(Y, q))$, (B_t^+, Y_t^+) is a $(n + 1)$ -dimensional Brownian motion, (Y_t^+, q_t^+) independent of (B, ξ) and Y_t^+ is $\sigma_s(Y_t^+, q_t^+)$ -Brownian motion, i.e. independent increments. Hence $\eta(t, \xi, B, q)$, B_t^+ and (Y_t^+, q_t^+) are independent under conditional probability $\dot{P}(\cdot/\sigma_t(Y, q))$, \dot{P} -almost surely, although the probability distribution of q_t^+ might depend on the past value of $(Y(\theta), q(\theta, A), \theta \leq t, A \in \mathcal{B}(I))$. Hence there exists a null set $N \in \sigma_t(Y, q)$, such that for $\omega \notin N$, $(\Omega, F, \dot{P}(\cdot/\sigma_t(Y, q))(\omega), \eta(t, \xi, B, q), B_t^+, Y_t^+, q_t^+) \in \mathfrak{X}$. Therefore

$$(6.20) \quad \dot{E}(\phi(\Lambda(s, Y_t^+, q_t^+, \Lambda(t, Y, q, \nu))/\sigma_t(Y, q))) \leq (S(s)\phi)(\Lambda(t, Y, q, \nu)),$$

\dot{P} -almost surely.

Combining (6.20) with (6.19), we have

$$(6.21) \quad J(t + s, \mathcal{A}, \nu, \phi) \leq S(t)(S(s)\phi)(\nu).$$

Taking the supremum with respect to $\mathcal{A} \in \mathfrak{X}(\nu/\|\nu\|)$, we have

$$(6.22) \quad S(t + s)\phi(\nu) \leq S(t)(S(s)\phi)(\nu).$$

For the converse inequality we will show some lemmas

LEMMA 1. *Let $N \subset (M(R^n), \Delta)$ be totally bounded. Then $\{\Lambda(t, Y_\omega, q_\omega, \nu); \mathcal{A} \in \mathfrak{X}(\nu/\|\nu\|), \nu \in N\}$ is totally bounded in (m, δ) .*

Proof. Consider the SDE, for $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, Y, q) \in \mathfrak{X}(\nu/\|\nu\|)$.

$$(6.23) \quad \begin{cases} d\eta(t) = \alpha(\eta(t))dB(t) + \int_r \gamma(\eta(t), u)\lambda(dt, du) \\ \eta(0) = \xi. \end{cases}$$

Then, using this unique solution $\eta(t) = \eta(t, \xi, B, \lambda)$ we have

$$(6.24) \quad \Lambda(t, y, \lambda, \nu)(A) = \|\nu\| \dot{E}\chi_A(\eta(t))\mathcal{L}(t, \xi, B, y, \lambda).$$

Hence, by (5.7)

$$(6.25) \quad \begin{aligned} \Lambda^2(t, y, \lambda, \nu)(A) &\leq \|\nu\|^2 \dot{P}(\gamma(t) \in A) \dot{E}^{\mathcal{L}^2}(t, \xi, B, y, \lambda) \\ &\leq \|\nu\|^2 \dot{P}(\gamma(t) \in A) \exp[2K_2(t+1)(\|y\|_t + 1)^2] \end{aligned}$$

where $\|y\|_t = \sup_{s \leq t} |y(s)|$. This means

$$(6.26) \quad \Lambda(t, y, \lambda, \nu)(A) \leq \|\nu\| \sqrt{\dot{P}(\gamma(t) \in A)} \exp[K_2(t+1)(1+a)^2],$$

whenever $\|y\|_t \leq a$.

On the other hand the condition (A1) implies that, for $\varepsilon' > 0$, there exists $b = b(\varepsilon', t, N)$ such that

$$(6.27) \quad \dot{P}(|\gamma(t, \xi, B, \lambda)| > b) < \varepsilon' \quad \text{for } \forall \lambda \in \hat{M}([0, T] \times \Gamma), \mathcal{A} \in \mathfrak{A}\left(\frac{\nu}{\|\nu\|}\right), \nu \in N.$$

Since Y is a Brownian motion, for $\varepsilon > 0$ there exists $a = a(\varepsilon)$ such that

$$(6.28) \quad \dot{P}(\sup_{s \leq t} |Y(s)| \leq a) > 1 - \varepsilon \quad \text{for } \forall \mathcal{A} \in \mathfrak{A}.$$

Putting $\varepsilon' = \varepsilon^2 e^{-2K_2(1+t)(1+a(\varepsilon))^2}$, (6.26) gives

$$(6.29) \quad \Lambda(t, y, \lambda, \nu)(K^c) < \varepsilon \|\nu\| \quad \text{for } \lambda \in \hat{M}([0, T] \times \Gamma)$$

whenever $\|y\|_t < a$, where the compact set K is given by

$$(6.30) \quad K = \{x \in R^n : |x| \leq b(\varepsilon', t, N)\}.$$

Therefore combining (6.28) and (6.29), we see, for $\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)$,

$$(6.31) \quad 1 - \varepsilon < \dot{P}(\sup_{s \leq t} |Y(s)| \leq a(\varepsilon)) \leq \dot{P}(\Lambda(t, Y, q, \nu)(K^c) < \varepsilon \|\nu\|).$$

Recalling the condition “ $0 < c' \leq \|\nu\| \leq c$ for $\forall \nu \in N$ ”, (6.31) implies Lemma 1 by virtue of Proposition 5.1.

Applying Prohorov's theorem, Lemma 1 gives

LEMMA 2. For $\varepsilon > 0$ and a totally bounded set $N \subset (M(R^n), \Delta)$ there exists a compact set $\tilde{N} = \tilde{N}(\varepsilon, t, N) \subset (M(R^n), \Delta)$ such that

$$(6.32) \quad \dot{P}(\Lambda(t, Y, q, \nu) \in \tilde{N}) > 1 - \varepsilon \quad \text{for } \mathcal{A} \in \mathfrak{A}\left(\frac{\nu}{\|\nu\|}\right), \nu \in N.$$

LEMMA 3. Suppose that $M(R^n) = M_0 \cup \dots \cup M_i$ is a Borel partition of $M(R^n)$. Let $\nu_i \in M_i$ and $\mathcal{A} = (\Omega_i, F_i, \dot{P}_i, \xi_i, B_i, Y_i, q_i) \in \mathfrak{A}(\nu_i/\|\nu_i\|)$. For any fixed $\mathcal{A} = (\Omega, F, P, \xi, B, Y, q) \in \mathfrak{A}(\nu/\|\nu\|)$ we define $\tilde{\Omega}, \tilde{F}, \tilde{P}, \tilde{\xi}, \tilde{B}, \tilde{Y}, \tilde{q}$ as follows.

$$\begin{aligned}\hat{\Omega} &= \Omega \times \Omega_0 \times \Omega_1 \times \cdots \times \Omega_\ell, & \tilde{F} &= F \times F_0 \times \cdots \times F_\ell \\ \hat{P} &= \dot{P} \times \dot{P}_1 \times \cdots \times \dot{P}_\ell, & \tilde{\xi} &= \xi \\ \tilde{B}(\theta) &= \begin{cases} B(\theta), & \theta \leq t \\ B(t) + \sum_{i=0}^{\ell} B_i(\theta - t)\chi_{M_i}(\Lambda(t, Y, q, \nu)), & \theta \geq t. \end{cases}\end{aligned}$$

\tilde{Y} is defined in the same way.

$$\tilde{q}(\theta, A) = \begin{cases} q(\theta, A), & \theta \leq t \\ q(t, A) + \sum_{i=0}^{\ell} q_i(\theta_i - t, A)\chi_{M_i}(\Lambda(t, Y, q, \nu)), & \theta \geq t. \end{cases}$$

Then $\tilde{\mathcal{A}} = (\tilde{\Omega}, \tilde{F}, \hat{P}, \tilde{\xi}, \tilde{B}, \tilde{Y}, \tilde{q}) \in \mathfrak{A}(\nu/\|\nu\|)$.

Proof. ξ is independent of $\{(B, Y, q), (\xi_i, B_i, Y_i, q_i), i = 0, \dots, \ell\}$. So $\tilde{\xi}$ is independent of $(\tilde{B}, \tilde{Y}, \tilde{q})$.

\tilde{B} is a Brownian motion, because for $g \in C_b((R^n)^k)$ and $\theta_j \geq t, j = 1, \dots, k$,

$$(6.33) \quad \begin{aligned} \hat{E}(g(\tilde{B}_j(\theta_j) - \tilde{B}(t), j = 1, \dots, k) / \sigma_i(B, Y, q) \vee \sigma(Y_i, q_i), i = 0, \dots, \ell) \\ = \hat{E}_i(g(B_i(\theta_j - t), j = 1, \dots, k) \quad \text{if } \Lambda(t, Y, q, \nu) \in M_i. \end{aligned}$$

Hence $(\tilde{B}(s) - \tilde{B}(t), s \geq t)$ is a Brownian motion which is independent of $\sigma_i(B, Y, q) \vee \sigma(Y_0, q_0) \vee \cdots \vee \sigma(Y_i, q_i)$, since B_i is a Brownian motion. This implies that $(\tilde{B}(s) - \tilde{B}(t), s \geq t, B(\theta), \theta \leq t)$ is independent of $((Y, q), (Y_i, q_i), i = 0, \dots, \ell)$, since B is independent of $((Y, q), (Y_i, q_i), i = 0, \dots, \ell)$. Therefore \tilde{B} is independent of (\tilde{Y}, \tilde{q}) , because (\tilde{Y}, \tilde{q}) is measurable with respect to $\sigma(Y, q, Y_0, q_0, \dots, Y_\ell, q_\ell)$.

Using a similar calculation as (6.33), we see that for $g \in C_b(R^k)$ and $\theta_j \geq \theta \geq t, j = 1, \dots, k$

$$(6.34) \quad \begin{aligned} \hat{E}(g(\tilde{Y}(\theta_j) - \tilde{Y}(\theta), j = 1, \dots, k) / \sigma_i(Y, q) \vee \sigma_{\theta-i}(Y_0, q_0, \dots, Y_i, q_i)) \\ = \hat{E}_i(g(Y_i(\theta_j) - Y_i(\theta), j = 1, \dots, k)), \quad \text{if } \Lambda(t, Y, q, \nu) \in M_i. \end{aligned}$$

Therefore $(\tilde{Y}(s) - \tilde{Y}(\theta), s \geq \theta)$ is a Brownian motion which is independent of $\sigma_i(\tilde{Y}, \tilde{q})$.

It is clear that \tilde{q} satisfies the conditions (v) and (vi), from the definition of \tilde{q} . This completes the proof of Lemma 3.

Now we prove the inequality (6.35) for Theorem.

$$(6.35) \quad S(t + s)\phi(\nu) \geq S(t)(S(s)\phi)(\nu).$$

Since $J(s, \pi, \nu, \phi)$ is continuous in π, ν and \mathcal{P} is compact, $J(s, \pi, \nu, \phi)$ is continuous in ν uniformly in $\pi \in \mathcal{P}$, namely for $\varepsilon > 0$ there exists $\delta(\varepsilon, \nu) = \delta(\varepsilon, \nu, s, \phi)$ such that, if $\Delta(\nu', \nu) > \delta(\varepsilon, \nu)$ then

$$(6.36) \quad |J(s, \pi, \nu', \phi) - J(s, \pi, \nu, \phi)| < \varepsilon.$$

Hence

$$(6.37) \quad |S(s)\phi(\nu') - S(s)\phi(\nu)| < \varepsilon, \quad \text{if } \Delta(\nu', \nu) < \delta(\varepsilon, \nu).$$

Applying Lemma 2 for $N = \{\nu\}$, given $\varepsilon > 0$ we can take a compact set \tilde{N} as in (6.32). From (6.36) and (6.37) we can take a Borel partition of \tilde{N} , say $\tilde{N} = M_1 \cup \dots \cup M_\ell$, such that, if $\nu', \nu'' \in M_i$, then

$$(6.38) \quad |J(s, \pi, \nu', \phi) - J(s, \pi, \nu'', \phi)| < \varepsilon$$

and

$$(6.39) \quad |S(s)\phi(\nu') - S(s)\phi(\nu'')| < \varepsilon.$$

Fix $\nu_i \in M_i$, $i = 1, \dots, \ell$, arbitrarily and take $\mathcal{A}_i \in \mathfrak{A}(\nu_i/\|\nu_i\|)$ such that

$$(6.40) \quad J(s, \pi_i, \nu_i, \phi) \geq S(s)\phi(\nu_i) - \varepsilon$$

where π_i is probability distribution of $(Y_{\mathcal{A}_i}, q_{\mathcal{A}_i})$. Then, by (6.38) ~ (6.40), we see

$$(6.41) \quad J(s, \pi_i, \nu', \phi) \geq J(s, \pi_i, \nu_i, \phi) - \varepsilon \geq S(s)\phi(\nu_i) - 2\varepsilon \geq S(s)\phi(\nu') - 3\varepsilon$$

for $\nu' \in M_i$.

Let $\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)$ and $M_0 = \tilde{N}^c$ and take $\nu_0 \in M_0(R^n)$ and $\mathcal{A}_0 \in \mathfrak{A}(\nu_0/\|\nu_0\|)$ arbitrarily. Then M_i , $i = 0, \dots, \ell$ is a Borel partition of $M(R^n)$. According to Lemma 3 we have $(\tilde{Q}, \tilde{F}, \tilde{P}, \tilde{\xi}, \tilde{B}, \tilde{Y}, \tilde{q}) \in \mathfrak{A}(\nu/\|\nu\|)$. Then

$$(6.42) \quad \begin{aligned} J(t+s, \tilde{\mathcal{A}}, \nu, \phi) &= \mathring{E}\phi(\Lambda(t+s, \tilde{Y}, \tilde{q}, \nu)) \\ &= \mathring{E}\phi(\Lambda(s, \tilde{Y}_t^+, \tilde{q}_t^+, \Lambda(t, Y, q, \nu))) \\ &= \mathring{E}[\mathring{E}(\phi(\Lambda(s, \tilde{Y}_t^+, \tilde{q}_t^+, \Lambda(t, Y, q, \nu)))/\sigma_t(Y, q))]. \end{aligned}$$

On the other hand, by (6.41) we have

$$\begin{aligned} &\mathring{E}(\phi(\Lambda(s, \tilde{Y}_t^+, \tilde{q}_t^+, \Lambda(t, Y, q, \nu)))/\sigma_t(Y, q)) \\ &= \sum_{i=0}^{\ell} [\mathring{E}_i\phi(\Lambda(s, Y_i, q_i, \Lambda(t, Y, q, \nu)))]\chi_{M_i}(\Lambda(t, Y, q, \nu)) \\ &= \sum_{i=0}^{\ell} J(s, \mathcal{A}_i, \Lambda(t, Y, q, \nu))\chi_{M_i}(\Lambda(t, Y, q, \nu)) \end{aligned}$$

$$\begin{aligned} &\geq \sum_{i=0}^t S(s)\phi(\Lambda(t, Y, q, \nu))\chi_{M_i}(\Lambda(t, Y, q, \nu)) - 3\varepsilon - \|\phi\|\chi_{M_0}(\Lambda(t, Y, q, \nu)) \\ &= S(s)\phi(\Lambda(t, Y, q, \nu)) - 3\varepsilon - \|\phi\|\chi_{M_0}(\Lambda(t, Y, q, \nu)). \end{aligned}$$

Combining with (6.42) we see

$$(6.43) \quad \begin{aligned} J(t+s, \tilde{\mathcal{A}}, \nu, \phi) &\geq \dot{E}(S(s)\phi)(\Lambda(t, Y, q, \nu)) - 3\varepsilon - \|\phi\|\dot{P}(\Lambda(t, Y, q, \nu) \notin \tilde{N}) \\ &\geq \dot{E}(S(s)\phi)(\Lambda(t, Y, q, \nu)) - 3\varepsilon - \|\phi\|\varepsilon. \end{aligned}$$

Since $\tilde{\mathcal{A}} \in \mathfrak{A}(\nu/\|\nu\|)$ we see

$$(6.44) \quad S(t+s)\phi(\nu) \geq \dot{E}(S(s)\phi)(\Lambda(t, Y, q, \nu)) - \varepsilon(3 + \|\phi\|).$$

Taking the supremum with respect to $\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)$, we conclude

$$(6.45) \quad S(t+s)\phi(\nu) \geq S(t)(S(s)\phi)(\nu) - \varepsilon(3 + \|\phi\|).$$

Tending $\varepsilon \downarrow 0$, we get our desired inequality (6.35). This completes the proof of Theorem 7.

§ 7. Generator and properties of $S(t)$.

We can easily see:

PROPOSITION 7.1. *The following properties hold,*

- (i) *monotone, $S(t)\phi \leq S(t)\psi$ whenever $\phi \leq \psi$*
- (ii) *contraction, $\|S(t)\phi - S(t)\psi\| \leq \|\phi - \psi\|$*
- (iii) *continuity, $S(\theta)\phi(\nu) \rightarrow S(t)\phi(\nu)$ as $\theta \rightarrow t$,*

uniformly on any compact set of $M(R^n)$.

That is, $S(t)$ is a monotone contraction weakly continuous semigroup on C .

Proof. (i) From the definition of J , (7.1) is clear

$$(7.1) \quad J(t, \pi, \nu, \phi) \leq J(t, \pi, \nu, \psi), \quad \text{if } \phi \leq \psi.$$

Hence taking the supremum with respect to $\pi \in \mathcal{P}$, we have (i).

$$(ii) \quad \begin{aligned} &|S(t)\phi(\nu) - S(t)\psi(\nu)| \\ &\leq \sup_{\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)} |\dot{E}\phi(\Lambda(t, Y_{\mathcal{A}}, q_{\mathcal{A}}, \nu)) - \dot{E}\psi(\Lambda(t, Y_{\mathcal{A}}, q_{\mathcal{A}}, \nu))| \\ &\leq \|\phi - \psi\|. \end{aligned}$$

Hence taking the supremum with respect to $\nu \in M(R^n)$, we have (ii).

(iii) By Proposition 6.1. $J(t, \pi, \nu, \phi)$ is continuous in $(t, \pi, \nu) \in [0, T] \times \mathcal{P} \times M(R^n)$. Since \mathcal{P} is compact, $S(t)\phi(\nu)$ is continuous in (t, ν) . Hence it is uniformly continuous on $[0, T] \times F$ where F is compact in $M(R^n)$. This implies (iii).

Now we calculate the generator of $S(t)$, according to [6]. We introduce the following set \mathcal{D} of functions ϕ which depend on finitely many scalar products. Fix $H_N \in C^\infty([0, \infty) \rightarrow [0, 1])$ such that $H_N(x) = 1$ on $[0, N]$, $= 0$ on $[N + 1, \infty)$ and decreasing in x .

$$(7.2) \quad \mathcal{D} = \{\phi; M(R^n) \longrightarrow R^1; \phi(\nu) = F(\langle f_1, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) H_N(\langle 1, \nu \rangle) \\ \text{with } F \in C_b^\infty(R^\ell), f_1, \dots, f_\ell \in C_0^\infty(R^n), \ell = 1, 2, \dots, N = 1, 2, \dots\}$$

where C_0^∞ denotes the space of C^∞ functions with compact supports and C_b^∞ the space of C^∞ functions with any bounded derivative. Clearly $\mathcal{D} \subset C$. Moreover we have

PROPOSITION 7.2. *For $\phi \in C$ there exists $\Phi_k \in \mathcal{D}$ such that $\Phi_k(\nu) \xrightarrow[k \rightarrow \infty]{} \phi(\nu)$ for any $\nu \in M(R^n)$.*

Proof. We can apply the same method as [6]. Let $v(x_i, 2^{-N})$ (= open ball with center x_i , radius 2^{-N}) $i = 1, 2, \dots$ be a covering of R^n with $\bigcup_{i=1}^{k_N} v(x_i, 2^{-N}) \supset [-N, N]^n$ (say I_N). Let g_ℓ^N , $\ell = 1, 2, \dots$ be a C^∞ -partition of unity such that

$$(7.3) \quad \text{supp } g_\ell^N \subset v(x_i, 2^{-N}) \text{ for some } i$$

$$(7.4) \quad \sum_{\ell=1}^{P_N} g_\ell = 1 \quad \text{on } I_N.$$

Take $\nu_i^N \in \text{supp } g_\ell^N \cap I_N$ arbitrarily. Putting $c_\ell^N(\nu) = \langle g_\ell^N, \nu \rangle$, we define ν_N by

$$(7.5) \quad \nu_N = \sum_{\ell=1}^{P_N} c_\ell^N(\nu) \delta_{\nu_i^N}.$$

Then $\|\nu\| \geq \|\nu_N\| \geq \nu(I_N)$ and for $g \in C_b(R^n)$

$$(7.6) \quad \langle g, \nu_N \rangle \rightarrow \langle g, \nu \rangle, \quad \text{as } N \rightarrow \infty.$$

Hence

$$(7.7) \quad \Delta(\nu_N, \nu) \rightarrow 0, \quad \text{as } N \rightarrow \infty.$$

Denote $F_N(z_1, \dots, z_{P_N}; \phi) = \phi(\sum_{\ell=1}^{P_N} z_\ell \delta_{\nu_i^N})$. That is,

$$(7.8) \quad \phi(\nu_N) = F_N(\langle g_1^N, \nu \rangle, \dots, \langle g_{P_N}^N, \nu \rangle; \phi)$$

Therefore, by (7.6), we have

$$(7.9) \quad \phi(\nu_N) \rightarrow \phi(\nu) \quad \text{as } N \rightarrow \infty.$$

From the definition of H_N , $\lim_{N \rightarrow \infty} H_N(\langle 1, \nu \rangle) = 1$ for any $\nu \in M(R^n)$. Hence

(7.9) gives

$$(7.10) \quad F_N(\langle g_1^N, \nu \rangle, \dots, \langle g_{p_N}^N, \nu \rangle; \phi) H_N(\langle 1, \nu \rangle) \rightarrow \phi(\nu), \quad \text{as } N \rightarrow \infty.$$

Take $\tilde{F}_N \in C_b^\infty(R^{p_N})$ such that

$$(7.11) \quad \|\tilde{F}_N - F_N\|_{L^\infty(I_{N+1})} < 2^{-N}.$$

Then we have

$$\begin{aligned} & |\tilde{F}_N(\langle g_1^N, \nu \rangle, \dots, \langle g_{p_N}^N, \nu \rangle) H_N(\langle 1, \nu \rangle) \\ & \quad - F_N(\langle g_1^N, \nu \rangle, \dots, \langle g_{p_N}^N, \nu \rangle; \phi) H_N(\langle 1, \nu \rangle)| < 2^{-N}. \end{aligned}$$

Combining with (7.10), we complete the proof.

We calculate the generator of $S(t)$, recalling (6.7). For an admissible switching system $\mathcal{A} \in \mathfrak{A}(\nu/\|\nu\|)$, we have the following Zakai equation for $\Lambda(t) = \Lambda(t, Y_{\mathcal{A}}, q_{\mathcal{A}}, \nu)$, (see Theorem 5.2 in [8]),

$$(7.12) \quad \begin{cases} d\langle f, \Lambda(t) \rangle = \langle A(U(t))f, \Lambda(t) \rangle dt + \langle hf, \Lambda(t) \rangle dY(t) \\ \langle f, \Lambda(0) \rangle = \langle f, \nu \rangle \quad \text{for } f \in C_b^2(R^n), \end{cases}$$

where

$$Y = Y_{\mathcal{A}}, \quad A(u) = \sum_{ij} a_{ij}(x) \frac{\partial^2}{\partial x_i \partial x_j} + \sum_i \gamma_i(x, y) \frac{\partial}{\partial x_i}$$

and $U(t)$ is a $\sigma_i(q)$ -progressively measurable process for q (see Theorem 4).

Therefore using a routine method we have

$$(7.13) \quad \dot{E}|\langle f, \Lambda(t) \rangle - \langle f, \nu \rangle|^2 \leq K_3(t+1) \int_0^t \dot{E}\langle 1, \Lambda(\theta) \rangle^2 d\theta$$

$$(7.14) \quad \dot{E}\langle 1, \Lambda(t) \rangle = \langle 1, \nu \rangle = \|\nu\|$$

and

$$(7.15) \quad \dot{E}\|\Lambda(t)\|^2 \leq \|\nu\|^2 e^{K_4 t}$$

where K_3 and K_4 are independent of \mathcal{A} . Combining these evaluations, we have

$$(7.16) \quad \dot{E}|\langle f, \Lambda(t) \rangle - \langle f, \nu \rangle|^2 \leq K_2 \|\nu\|^2 (t+1) t e^{K_4 t}.$$

Let $\Phi \in \mathcal{D}$, say $\Phi(\nu) = F(\langle f_1, \nu \rangle, \dots, \langle f_l, \nu \rangle) H_N(\langle 1, \nu \rangle)$. For simplicity we put $f_0 = 1$, and $\tilde{F}(z_0, z_1, \dots, z_l) = F(z_1, \dots, z_l) H_N(z_0)$. Appealing to Ito's formula, we see

$$\begin{aligned}
d\Phi(\Lambda(t)) &= \sum_{i=0}^{\ell} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \Lambda(t) \rangle, \dots, \langle f_\ell, \Lambda(t) \rangle) \langle A(U(t))f_i, \Lambda(t) \rangle dt \\
(7.17) \quad &+ \frac{1}{2} \sum_{i,j=0}^{\ell} \frac{\partial^2 \tilde{F}}{\partial z_i \partial z_j} (\langle f_0, \Lambda(t) \rangle, \dots, \langle f_\ell, \Lambda(t) \rangle) \langle hf_i, \Lambda(t) \rangle \langle hf_j, \Lambda(t) \rangle dt \\
&+ \sum_{i=0}^{\ell} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \Lambda(t) \rangle, \dots, \langle f_\ell, \Lambda(t) \rangle) \langle hf_i, \Lambda(t) \rangle dY(t).
\end{aligned}$$

Using (7.16) we have, for $s \leq 1$,

$$\begin{aligned}
(7.18) \quad &\int_0^s \dot{E} \left| \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \Lambda(t) \rangle, \dots, \langle f_\ell, \Lambda(t) \rangle) \langle A(U(t))f_i, \Lambda(t) \rangle \right. \\
&\quad \left. - \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle A(U(t))f_i, \nu \rangle \right| dt \leq K_5 \|\nu\| s^{3/2}
\end{aligned}$$

and

$$\begin{aligned}
(7.19) \quad &\int_0^s \dot{E} \left| \frac{\partial^2 \tilde{F}}{\partial z_i \partial z_j} (\langle f_0, \Lambda(t) \rangle, \dots, \langle f_\ell, \Lambda(t) \rangle) \langle hf_i, \Lambda(t) \rangle \langle hf_j, \Lambda(t) \rangle \right. \\
&\quad \left. - \frac{\partial^2 \tilde{F}}{\partial z_i \partial z_j} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle hf_i, \nu \rangle \langle hf_j, \nu \rangle \right| dt \leq K_5 \|\nu\| s^{3/2}
\end{aligned}$$

with K_5 which is independent of \mathcal{A} and ν .

Define $G\Phi$ as follows

$$\begin{aligned}
(7.20) \quad G\Phi(\nu) &= \sup_{u \in \Gamma} \left(\sum_{i=0}^{\ell} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle A(u)f_i, \nu \rangle \right) \\
&+ \frac{1}{2} \sum_{i,j=0}^{\ell} \frac{\partial^2 \tilde{F}}{\partial z_i \partial z_j} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle hf_i, \nu \rangle \langle hf_j, \nu \rangle.
\end{aligned}$$

Since \tilde{F} is smooth and $\langle A(u)f_i, \nu \rangle$ continuous in u and ν , $G\Phi(\nu)$ is continuous in ν . Moreover $G\Phi(\nu) = 0$ whenever $\|\nu\| \geq N + 1$. Therefore $G\Phi$ is bounded. This implies that $G\Phi \in C$ for $\Phi \in \mathcal{D}$.

We remark that

$$\begin{aligned}
(7.21) \quad &\sum_i \int_0^s \dot{E} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle A(U(t))f_i, \nu \rangle dt \\
&\leq \int_0^s \sup_{u \in \Gamma} \left(\sum_{i=0}^{\ell} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle A(u)f_i, \nu \rangle \right) dt \\
&= s, \sup_{u \in \Gamma} \left(\sum_{i=0}^{\ell} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle A(u)f_i, \nu \rangle \right) \\
&= \sup_{u \in \Gamma} \sum_{i=0}^{\ell} \int_0^s \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle A(u)f_i, \nu \rangle dt \\
&\leq \sup_{U; \text{usual control for switch relaxed syst.}} \sum_{i=0}^{\ell} \int_0^s \dot{E} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_\ell, \nu \rangle) \langle A(U(t))f_i, \nu \rangle dt.
\end{aligned}$$

Taking the supremum of left side of (7.21) with respect to $U(t)$ for switching admissible systems, we have

$$(7.22) \quad \frac{1}{s} \sup_{U; \text{ switch syst.}} \sum_{i=0}^{\ell} \int_0^s \dot{E} \frac{\partial \tilde{F}}{\partial z_i} (\langle f_0, \nu \rangle, \dots, \langle f_i, \nu \rangle) \langle A(U(t))f_i, \nu \rangle dt = G\Phi(\nu).$$

On the other hand (7.17) ~ (7.19) tell us that

$$(7.23) \quad \begin{aligned} & \left| \dot{E}\Phi(\Lambda(s, Y_{\mathcal{A}}, q_{\mathcal{A}}, \nu)) - \Phi(\nu) - \int_0^s \sum_{i=0}^{\ell} \dot{E} \frac{\partial \tilde{F}}{\partial z_j} (\langle f_0, \nu \rangle, \dots, \langle f_i, \nu \rangle) \langle A(U(t))f_i, \nu \rangle \right. \\ & \quad \left. + \frac{1}{2} \sum_{i,j=0}^{\ell} \frac{\partial^2 \tilde{F}}{\partial z_i \partial z_j} (\langle f_0, \nu \rangle, \dots, \langle f_i, \nu \rangle) \langle hf_i, \nu \rangle \langle hf_j, \nu \rangle dt \right| \\ & \leq 2K_s(n+1)^2 \|\nu\| s^{3/2}. \end{aligned}$$

Appealing to (7.22) we have

$$(7.24) \quad \left| \frac{1}{s} (S(s)\Phi(\nu) - \Phi(\nu)) - G\Phi(\nu) \right| \leq 2K_s(n+1)^2 \|\nu\| s^{1/2}$$

Recalling (5.16) we have

$$(7.25) \quad \|\Lambda(s, Y_{\mathcal{A}}, q_{\mathcal{A}}, \nu)\| \geq \|\nu\| e^{-K_1(a+1)^2(s+1)} \quad \text{whenever } \sup_{t \leq s} |Y_{\mathcal{A}}(t)| \leq a.$$

Since $Y_{\mathcal{A}}$ is a Brownian motion, a martingale inequality implies

$$(7.26) \quad \dot{P}(\sup_{t \leq s} |Y_{\mathcal{A}}(t)| \geq a) \leq \frac{s}{a^2}.$$

Putting $a = 1/\sqrt{\varepsilon}$ and $\tilde{N}(\varepsilon) = (N+1)e^{K_1(a+1)^2(s+1)}$, we see, from (7.25) and (7.26)

$$(7.27) \quad \begin{aligned} \dot{P}(\|\Lambda(s, Y_{\mathcal{A}}, q_{\mathcal{A}}, \nu)\| \leq N+1) & \leq \dot{P}\left(\sup_{t \leq s} |Y_{\mathcal{A}}(t)| > \frac{1}{\sqrt{\varepsilon}}\right) < \varepsilon s \\ & \text{whenever } \|\nu\| \geq \tilde{N}(\varepsilon). \end{aligned}$$

Therefore, if $\|\nu\| \geq \tilde{N}(\varepsilon)$, then

$$(7.28) \quad E|\Phi(\Lambda(s, Y_{\mathcal{A}}, q_{\mathcal{A}}, \nu))| \leq \|\Phi\| \varepsilon s.$$

This implies, by virtue of " $\Phi(\nu) = G\Phi(\nu) = 0$ for $\|\nu\| \geq \tilde{N}(\varepsilon)$ ",

$$(7.29) \quad \frac{1}{s} (S(s)\Phi(\nu) - \Phi(\nu)) - G\Phi(\nu) \leq \|\Phi\| \varepsilon, \quad \text{whenever } \|\nu\| \geq \tilde{N}(\varepsilon).$$

Appealing to (7.24), we have

$$(7.30) \quad \left| \frac{1}{s} (S(s)\Phi(\nu) - \Phi(\nu)) - G\Phi(\nu) \right| \leq (1 + \|\Phi\|)\varepsilon,$$

whenever $s \leq \varepsilon / (2K_\varepsilon(n + 1)^2 \tilde{N}(\varepsilon))^2$.

This implies that $1/s(S(s)\Phi(\nu) - \Phi(\nu))$ converges to $G\Phi(\nu)$ uniformly on $M(R^n)$, as $s \downarrow 0$.

THEOREM 8. $\mathcal{D}(\mathfrak{G}) \supset \mathcal{D}$ and

$$(7.31) \quad \mathfrak{G}\Phi = G\Phi \quad \text{on } \mathcal{D}$$

that is, the generator is an extension of G .

§ 8. Time discrete approximation

First we recall an approximation theorem of Proposition 4.1 and Theorem 4, namely, for $\mathcal{A} = (\Omega, F, \dot{P}, \xi, B, T, q)$ there exists an approximate usual control U_n , such that

$$(8.1) \quad U_n \text{ is an } \sigma_t(q)\text{-progressively measurable } \Gamma\text{-valued process}$$

$$(8.2) \quad q_n(t, A) = \int_0^t \delta_{U_n(s)}(A) ds \text{ is a switching relaxed control of } \mathcal{A}_n = (\Omega, F, \dot{P}, \xi, B, Y, q_n)$$

and

$$(8.3) \quad J(t, \mathcal{A}, \nu, \phi) = \lim_{n \rightarrow \infty} J(t, \mathcal{A}_n, \nu, \phi) \text{ for } \forall \phi \in C.$$

Now we define a usual admissible system $\tilde{\mathcal{A}} = (\Omega, F, \dot{P}, \xi, B, Y, U)$ as follows: $(\Omega, F, \dot{P}, \xi, B, Y)$ satisfies the same conditions as an admissible (relaxed) system, U is a Γ -valued process, ξ, B and (Y, U) are independent and Y is a $\sigma_t(Y, U)$ -Brownian motion.

\mathfrak{U} denotes the totality of usual admissible systems, and we apply similar notations as for the relaxed case. Putting $\tilde{q}_U(t, A) = \int_0^t \delta_{U(s, \omega)}(A) ds$, we see $\mathcal{A}_U = (\Omega, F, P, \xi, B, Y, \tilde{q}_U) \in \mathfrak{U}$. Thus a usual admissible system can be regarded as an admissible (relaxed) system. Moreover a unique solution \tilde{X} of the SDE,

$$(8.4) \quad \begin{cases} d\tilde{X}(t) = \alpha(\tilde{X}(t))dB(t) + \gamma((t), U(t))dt \\ \tilde{X}(0) = \xi \end{cases}$$

gives a unique solution $X_{\mathcal{A}_U} (= \hat{X})$.

Since (ξ, B) and (Y, U) are independent, we can calculate

$$\dot{E}(f(\hat{X}(\theta))L(\theta, \tilde{\mathcal{A}})/\sigma_t(Y, U))$$

in the same way as (5.8), and get

$$(8.5) \quad \begin{aligned} & \dot{E}(f(\hat{X}(\theta))L(\theta, \tilde{\mathcal{A}})/Y = y, U(t) = v(t) \text{ a.a. } t) \\ &= \dot{E}f(\gamma_v(\theta))\mathcal{L}(\theta, \xi, B, y, q_v) \\ &= \dot{E}(f(X_{\mathcal{A}_U}(\theta))L(\theta, \mathcal{A}_U)/Y = y, q_U = \lambda_v) \end{aligned}$$

where $\lambda_v(t, A) = \int_0^t \delta_{v(s)}(A) ds$. We remark that, if $v(t) = v^*(t)$, a.a. t , then $\lambda_v = \lambda_{v^*}$. The unnormalized conditional distribution $\tilde{\lambda}(t, \tilde{\mathcal{A}}, v)$ is defined by

$$(8.6) \quad \langle f, \tilde{\lambda}(\theta, \tilde{\mathcal{A}}, v) \rangle = \dot{E}_v \langle f(\hat{X}(\theta))L(\theta, \tilde{\mathcal{A}})/\sigma_t(Y, U) \rangle \quad \text{for } f \in C_b(\mathbb{R}^n).$$

Hence (8.5) implies

$$(8.7) \quad \begin{aligned} \tilde{\lambda}(\theta, \tilde{\mathcal{A}}, v) &= \lambda(\theta, Y, q_U, \nu), \quad \dot{P}\text{-almost surely.} \\ \tilde{\lambda}(\theta, \tilde{\mathcal{A}}, \nu) &= \tilde{\lambda}(\theta, Y, U, \nu) \quad \text{and} \quad \tilde{J}(\theta, \tilde{\mathcal{A}}, \nu, \phi) = \dot{E}\phi(\tilde{\lambda}(\theta, \tilde{\mathcal{A}}, \nu)) = \tilde{J}(\theta, Y, U, \nu, \phi). \end{aligned}$$

We approximate U_n of (8.1) (say W for simplicity) by a switching usual control, by a routine method, i.e. \tilde{U}_k and $\tilde{U}_{k,p}$ are defined as follows

$$(8.7) \quad \tilde{U}_k(t) = 2^k \int_{t-2^{-k}}^t W(s) ds \quad \text{and} \quad \tilde{U}_{k,p}(t) = \tilde{U}_k\left(\frac{[2^p t]}{2^p}\right).$$

Then $\lim_{k \rightarrow \infty} \lim_{p \rightarrow \infty} \tilde{U}_{k,p}(t) = W$ in $L^2([0, T] \times \Omega)$. This fact implies that there exists an approximate switching usual control W_k , which is $\sigma_t(U)$ -progressively measurable and satisfies

$$(8.8) \quad \dot{E} \int_0^T |W_k(s) - W(s)|^2 ds \rightarrow 0, \quad \text{as } k \rightarrow \infty$$

and

$$(8.9) \quad \tilde{J}(t, Y, W, \nu, \phi) = \lim_{k \rightarrow \infty} \tilde{J}(t, Y, W_k, \nu, \phi).$$

By (8.8) some subsequence of W_k converges to W a.e. in $[0, T] \times \Omega$, we assume " $W_k \rightarrow W$ a.e." for simplicity. Therefore, for a.a. $\omega(\dot{P})$,

$$(8.10) \quad \begin{aligned} \int_{\Gamma} g(u) \tilde{q}_{W_k}(t, du) &= \int_0^t \int_{\Gamma} g(u) \delta_{W_k}(du) ds \\ &= \int_0^t g(W_k(s)) ds \xrightarrow[k \rightarrow \infty]{} \int_0^t g(W(s)) ds = \int_{\Gamma} g(u) \tilde{q}_W(t, du) \end{aligned}$$

for any t and $g \in C_b(\Gamma)$.

This implies, as $k \rightarrow \infty$.

$$(8.11) \quad \hat{d}_T(\tilde{q}_{W_k}, \tilde{q}_W) \rightarrow 0, \quad \dot{P}\text{-almost surely.}$$

Hence by Propositions 2.2 and the Corollary of Proposition 3.1 we see

$$(8.12) \quad \mathcal{A}_{W_k} \rightarrow \mathcal{A}_W \quad \text{and} \quad (X_{W_k}, \mathcal{A}_{W_k}) \rightarrow (X_W, \mathcal{A}_W) \quad \text{in } D_T.$$

Therefore Theorem 6 implies that, for $\forall \theta$,

$$(8.13) \quad \tilde{\Lambda}(\theta, Y, W_k, \nu) \rightarrow \tilde{\Lambda}(\theta, Y, W, \nu) \quad \text{in metric } \delta.$$

So we have

PROPOSITION 8.1. *For $\mathcal{A} = (\Omega, F, P, \xi, B, Y, q)$ there exists a switching usual control $W_k, k = 1, 2, \dots$, which is $\sigma_t(q)$ -progressively measurable and for any θ*

$$\tilde{\Lambda}(\theta, Y, W_k, \nu) \rightarrow \Lambda(\theta, Y, q, \nu) \quad \text{in metric } \delta, \text{ as } k \rightarrow \infty.$$

This means that switching usual controls are rich enough in the class of relaxed controls.

Put $\tilde{\mathfrak{A}}_N =$ totality of usual admissible systems whose usual controls are switching with time interval 2^{-N} , i.e. $\tilde{\mathcal{A}} = (\Omega, F, \dot{P}, \xi, B, Y, U) \in \tilde{\mathfrak{A}}_N$, iff $U(t) = U([\cdot]2^N)$. We denote $\tilde{\mathcal{A}} \in \tilde{\mathfrak{A}}_0$, if U is constant control. So $\tilde{\mathfrak{A}}_0 \subset \tilde{\mathfrak{A}}_N$. When $P_\xi = \nu/\|\nu\|$, we say $\tilde{\mathcal{A}} \in \tilde{\mathfrak{A}}_N(\nu/\|\nu\|)$. Put $\tilde{\mathfrak{A}} = \bigcup_{N=0}^{\infty} \tilde{\mathfrak{A}}_N$.

From Proposition 8.1 we see

$$(8.14) \quad \begin{aligned} S(t)\phi(\nu) &= \sup_{\tilde{\mathcal{A}} \in \tilde{\mathfrak{A}}(\nu/\|\nu\|)} \tilde{J}(t, \tilde{\mathcal{A}}, \nu, \phi) \\ &= \lim_{N \rightarrow \infty} \sup_{\tilde{\mathcal{A}} \in \tilde{\mathfrak{A}}_N(\nu/\|\nu\|)} \tilde{J}(t, \tilde{\mathcal{A}}, \nu, \phi). \end{aligned}$$

Define $Q = Q_N$ by

$$(8.15) \quad Q\phi(\nu) = \sup_{\tilde{\mathcal{A}} \in \tilde{\mathfrak{A}}_N(\nu/\|\nu\|)} J(2^{-N}, \tilde{\mathcal{A}}, \nu, \phi).$$

We remark that

$$(8.16) \quad \begin{aligned} J(2^{-N}, \tilde{\mathcal{A}}, \nu, \phi) &= \dot{E}\phi(\tilde{\Lambda}(2^{-N}, Y, U(0), \nu)) \\ &= \int_r \dot{E}(\phi(\tilde{\Lambda}(2^{-N}, Y, U(0), \nu))/U(0) = u)P_{U_0}(du). \end{aligned}$$

Since $(Y(\theta), \theta \geq 0)$ is independent of $U(0)$,

$$(8.17) \quad \dot{E}(\phi(\tilde{\Lambda}(2^{-N}, Y, U(0), \nu/U(0) = u))) = \tilde{J}(2^{-N}, Y, u, \nu, \phi).$$

Moreover the value of the left side depends only on N, u, ν, ϕ , since Y is a Brownian motion with respect to $\dot{P}(\cdot/U(0) = u)$. We put

$$(8.18) \quad \tilde{J}(2^{-N}, Y, u, \nu, \phi) = \Phi_N(u, \nu; \phi)$$

and

$$(8.19) \quad \tilde{Q}\phi(\nu) = \sup_{u \in \Gamma} \Phi_N(u, \nu, \phi).$$

Then $\Phi_N(u, \nu, \phi)$ is continuous in $(u, \nu) \in \Gamma \times (M(R^n), \mathcal{A})$ and $\tilde{J}(2^{-N}, Y, u, \nu, \phi) \leq \tilde{Q}\phi(\nu)$. Combining with (8.12), we see

$$(8.20) \quad \begin{aligned} Q\phi(\nu) &\leq \tilde{Q}\phi(\nu) = \sup_{\tilde{\mathcal{A}} \in \mathfrak{A}_0(\nu/\|\nu\|)} \tilde{J}(2^{-N}, \tilde{\mathcal{A}}, \nu, \phi) \\ &\leq \sup_{\tilde{\mathcal{A}} \in \mathfrak{A}_N(\nu/\|\nu\|)} \tilde{J}(2^{-N}, \tilde{\mathcal{A}}, \nu, \phi) = Q\phi(\nu). \end{aligned}$$

Hence we have

$$(8.21) \quad Q_N = \tilde{Q}.$$

Since Γ is compact, using a measurable selection theorem, we can take a Borel measurable mapping, $v = v_\phi: M(R^n) \rightarrow \Gamma$ such that

$$(8.22) \quad \Phi_N(v_\phi(\nu), \nu, \phi) = \sup_{u \in \Gamma} \Phi_N(u, \nu, \phi).$$

This gives

$$(8.23) \quad Q\phi(\nu) = \tilde{J}(2^{-N}, Y, v_\phi(\nu), \nu, \phi)$$

We have, for $\tilde{\mathcal{A}} \in \mathfrak{A}_N(\nu/\|\nu\|)$

$$(8.24) \quad \begin{aligned} J(2^{-N+1}, \tilde{\mathcal{A}}, \nu, \phi) &= \dot{E}\phi(\tilde{\Lambda}(2^{-N+1}, Y, U, \nu)) \\ &= \dot{E}\phi(\tilde{\Lambda}(2^{-N}, Y_{2^{-N}}^+, U_{2^{-N}}^+, \tilde{\Lambda}(2^{-N}, Y, U, \nu))) \\ &\leq (Q\phi)(\tilde{\Lambda}(2^{-N}, Y, U, \nu)) \leq Q(Q\phi)(\nu) \\ &= Q^2\phi(\nu). \end{aligned}$$

Define $U_2 = U_{2, \phi, \nu}$ by

$$(8.25) \quad U_2(t) = \begin{cases} v_{Q\phi}(\nu), & 0 \leq t \leq 2^{-N} \\ v_\phi(\tilde{\Lambda}(2^{-N}, Y, v_{Q\phi}(\nu), \nu)), & 2^{-N} < t. \end{cases}$$

Then $\mathcal{A}^* = (\Omega, F, \dot{P}, \xi, B, Y, U_2) \in \mathfrak{A}_N(\nu/\|\nu\|)$ and

$$(8.26) \quad \begin{aligned} J(2^{-N+1}, \mathcal{A}^*, \nu, \phi) &= \dot{E}\phi(\tilde{\Lambda}(2^{-N+1}, Y, U_2, \nu)) \\ &= \dot{E}\dot{E}(\phi(\tilde{\Lambda}(2^{-N}, Y_{2^{-N}}^+, v_\phi(\tilde{\Lambda}(2^{-N}, Y, v_{Q\phi}(\nu), \nu))), \\ &\quad \tilde{\Lambda}(2^{-N}, Y, v_{Q\phi}(\nu), \nu))/\sigma_{2^{-N}}(Y)) \\ &= \dot{E}(Q\phi)(\tilde{\Lambda}(2^{-N}, Y, v_{Q\phi}(\nu), \nu)) \\ &= Q(Q\phi)(\nu) = Q^2\phi(\nu). \end{aligned}$$

Combining with (8.24), we have

$$(8.27) \quad \begin{aligned} Q^2\phi(\nu) &= \sup_{\mathcal{A} \in \mathfrak{A}_N(\nu/\|\nu\|)} \tilde{J}(2^{-N+1}, \mathcal{A}, \nu, \phi) \\ &= \tilde{J}(2^{-N+1}, \mathcal{A}^*, \nu, \phi). \end{aligned}$$

Repeating a similar calculation we see

$$(8.28) \quad Q^{k+1}\phi(\nu) = \sup_{\mathcal{A} \in \mathfrak{A}_N(\nu/\|\nu\|)} \tilde{J}((k+1)2^{-N}, \mathcal{A}, \nu, \phi)$$

and an optimal one $U_{k+1} = U_{k+1, \nu, \phi}$ is given successively by

$$U_{k+1}(t) = \begin{cases} U_{k, \nu, Q^k\phi}(t), & 0 \leq t \leq k2^{-N} \\ v_\phi(\tilde{A}(k2^{-N}, Y, U_{k, \nu, Q^k\phi}, \nu), & k2^{-N} < t. \end{cases}$$

Recalling (8.14) we see, for binary t (say $j2^{-p}$)

$$\lim_{N \rightarrow \infty} \sup_{\mathcal{A} \in \mathfrak{A}_N(\nu/\|\nu\|)} \tilde{J}(t, \mathcal{A}, \nu, \phi) = S(t)\phi(\nu) = \lim_{N \rightarrow \infty} Q_N^{[2^N t]}\phi(\nu)$$

and an approximate optimal usual switching control is given by $U_{k, \nu, \phi}$.

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