By using n-dimensional Levy's theorem i.e., Theorem 3.12 we arrive at the following multi-dimensional Girsanov's theorem.

Theorem 3.20 (Girsanov's theorem, n-dimensional). Let  $\mathbf{B}(t) = (B_1(t), B_2(t), \dots, B_n(t))$ :  $0 \le t \le T$  be a n-dimensional Brownian motion associated with a given filtration  $\{\mathcal{F}_t\}$ . Let  $\mathbf{\Theta}(t) = (\theta_1(t), \dots, \theta_n(t))$  be a n-dimensional adapted process such that

$$Z(t) = \exp\left\{-\int_0^t \sum_{i=1}^n \theta_i(s) \, dB_i(s) - \frac{1}{2} \int_0^t \sum_{i=1}^d \theta_i^2(s) \, ds\right\}$$

becomes a martingale. Then the stochastic process, defined by

$$\bar{\mathbf{B}}(t) = \mathbf{B}(t) + \int_0^t \mathbf{\Theta}(s) \, ds \quad 0 \le t \le T$$

is a n-dimensional Brownian motion under the probability measure  $\tilde{\mathbb{P}}$ , where

$$\tilde{\mathbb{P}}(A) = \int_A Z(T) d\mathbb{P}, \quad \forall A \in \mathcal{F}.$$

We have n-dimensional version of Corollary 3.19.

Corollary 3.21. Let  $\mathbf{B}(t) = (B_1(t), B_2(t), \dots, B_m(t)) : 0 \le t \le T$  be a m-dimensional Brownian motion relative to a filtration  $\{\mathcal{F}_t\}$ , and  $\mathbf{b} = (b_1, b_2, \dots, b_n) \in \mathbb{L}_n^1[0, T]$ ,  $\mathbf{\Theta}(t) = ((\theta_{ij})) \in \mathcal{Y}_{n \times m}(0, T)$ . Let  $\mathbf{Y}(t) = (Y_1(t), Y_2(t), \dots, Y_n(t))$  be a n-dimensional Ito process of the form

$$d\mathbf{Y}(t) = \mathbf{b}(t) dt + \mathbf{\Theta}(t) d\mathbf{B}(t), \quad 0 \le t \le T.$$

Suppose there exist  $\mathbf{u} = (u_1, u_2, \dots, u_m) \in \mathcal{Y}_{1 \times m}(0, T)$  and  $\mathbf{a} \in \mathbb{L}^1_n(0, T)$  such that

$$\boldsymbol{\Theta}(t)\boldsymbol{u}(t) = \mathbf{b}(t) - \mathbf{a}(t)$$

and

$$Z(t) = \exp\left\{-\int_0^t \sum_{i=1}^m u_i(s) dB_i(s) - \frac{1}{2} \int_0^t \sum_{i=1}^m u_i^2(s) ds\right\}$$

becomes a martingale. Then the stochastic process, defined by

$$\bar{\mathbf{B}}(t) = \mathbf{B}(t) + \int_0^t \mathbf{u}(s) \, ds \quad 0 \le t \le T$$

is a m-dimensional Brownian motion under the probability measure  $\tilde{\mathbb{P}}$ , where

$$\tilde{\mathbb{P}}(A) = \int_A Z(T) d\mathbb{P}, \quad \forall A \in \mathcal{F}.$$

Moreover, in terms of  $\bar{\mathbf{B}}(t)$ , the n-dimensional process  $\mathbf{Y}(t)$  has the following representation:

$$d\mathbf{Y}(t) = \mathbf{a}(t) dt + \mathbf{\Theta}(t) d\bar{\mathbf{B}}(t).$$

**Example 3.15.** Consider a 2-dimensional Ito process  $\mathbf{Y}(t) = (Y_1(t), Y_2(t))$  given by

$$dY_1(t) = 2dt + dB_1(t) + dB_2(t), \quad dY_2(t) = 4dt + dB_1(t) - dB_2(t)$$

where  $\mathbf{B}(t) = (B_1(t), B_2(t))$  is a 2-dimensional Brownian motion. Find a probability measure  $\bar{\mathbb{P}}$  such that  $\mathbf{Y}(t)$  is a martingale with respect to  $\bar{\mathbb{P}}$ .

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Solution: Given Ito process can be written in the form

$$d\mathbf{Y}(t) = \mathbf{b}(t) dt + \mathbf{\Theta}(t) d\mathbf{B}(t), \quad 0 \le t \le T.$$

where  $\mathbf{b} = (2,4) \in \mathbb{L}_2^1[0,T]$  and  $\mathbf{\Theta}(t) = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \in \mathcal{Y}_{2\times 2}(0,T)$ . Let  $\mathbf{u} = \begin{pmatrix} u_1, u_2 \end{pmatrix} \in \mathcal{Y}_{1\times 2}(0,T)$  such that  $\mathbf{\Theta}(t)\mathbf{u}(t) = \mathbf{b}(t)$ . Then  $u_1(t) = 3$  and  $u_2(t) = -1$ . Note that the process

$$Z(t) = \exp\left\{-\int_0^t \sum_{i=1}^2 u_i(s) dB_i(s) - \frac{1}{2} \int_0^t \sum_{i=1}^2 u_i^2(s) ds\right\}$$

becomes a martingale (Novikov condition holds trivially). Define a probability measure  $\bar{\mathbb{P}}$  as

$$\bar{\mathbb{P}}(A) = \int_A Z(T) d\mathbb{P}, \quad \forall A \in \mathcal{F}.$$

Then by Corollary 3.21, the stochastic process

$$\bar{\mathbf{B}}(t) = \mathbf{B}(t) + \int_0^t \mathbf{u}(s) \, ds \quad 0 \le t \le T$$

is a 2-dimensional Brownian motion under the probability measure  $\bar{\mathbb{P}}$ . Moreover,  $\mathbf{Y}(t)$  can be written as

$$d\mathbf{Y}(t) = \mathbf{\Theta}(t) \, d\bar{\mathbf{B}}(t).$$

This shows that  $\mathbf{Y}(t)$  is a martingale with respect to the probability measure  $\bar{\mathbb{P}}$ .

## 4. STOCHASTIC DIFFERENTIAL EQUATION AND CONNECTION WITH PDES

Let B(t) be a m-dimensional Brownian motion and Z be a random variable independent of  $B(\cdot)$ . Consider a stochastic differential equation (**SDE**) of the type

$$\begin{cases} dX(t) = a(t, X(t)) dt + \sigma(t, X(t)) dB(t) \\ X(0) = Z \end{cases}$$

$$(4.1)$$

where  $a:[0,T]\times\mathbb{R}^n\to\mathbb{R}^n$  and  $\sigma:[0,T]\times\mathbb{R}^n\to\mathbb{R}^{n\times m}$  are given functions. We now discuss existence and uniqueness of solution of the SDE. Let  $\mathcal{F}_t$  be the filtration generated by Z and B(t) i.e.,  $\mathcal{F}_t=\sigma(Z,B(s):0\leq s\leq t)$ . Let us first define the solution concept.

**Definition 4.1.** Let B(t) be a m-dimensional Brownian motion on a given probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  and Z be a random variable on it, independent of  $B(\cdot)$ . An  $\mathbb{R}^n$ -valued stochastic process  $X(\cdot)$  on  $(\Omega, \mathcal{F})$  is a (strong) solution of the SDE (4.1) if

- i) X(t) is  $\mathcal{F}_t$ -adapted stochastic process where  $\mathcal{F}_t = \sigma(Z, B(s) : 0 \le s \le t)$ .
- ii) For all  $0 \le t \le T$ , there holds

$$X(t) = Z + \int_0^t a(s, X(s)) \, ds + \int_0^t \sigma(s, X(s)) \, dB(s)$$

where the integrals on the right hand side are well-defined.

We now prove a technical lemma so called **Gronwall's lemma** which is useful in our uniqueness proof.

**Lemma 4.1.** Let  $\phi(\cdot)$  and  $f(\cdot)$  are nonnegative continuous function on [0,T] and  $C \ge 0$  be a constant. If  $\phi(t) \le C + \int_0^t f(s)\phi(s) \, ds$  for all  $t \in [0,T]$ , then

$$\phi(t) \le Ce^{\int_0^t f(s) \, ds}, \quad t \in [0, T].$$

*Proof.* Let  $h(t) = C + \int_0^t f(s)\phi(s) ds$ . Then by given condition,  $\phi(t) \leq h(t)$  and hence, since f is nonnegative,  $f(t)\phi(t) \leq f(t)h(t)$ . We have

$$h'(t) = f(t)\phi(t) \le f(t)h(t)$$

$$\implies (h'(t) - h(t)f(t))e^{-\int_0^t f(s) \, ds} \le 0$$

$$\implies (h(t)e^{-\int_0^t f(s) \, ds})' \le 0$$

$$\implies h(t)e^{-\int_0^t f(s) \, ds} \le h(0)e^{-\int_0^0 f(s) \, ds} = C$$

$$\implies h(t) \le Ce^{\int_0^t f(s) \, ds}$$

$$\implies \phi(t) \le Ce^{\int_0^t f(s) \, ds}.$$

This completes the proof.

We now prove existence and uniqueness theorem for SDE.

**Theorem 4.2.** Let  $a:[0,T]\times\mathbb{R}^n\to\mathbb{R}^n,\ \sigma:[0,T]\times\mathbb{R}^n\to\mathbb{R}^{n\times m}$  be measurable functions satisfying

$$|a(t,x)| + |\sigma(t,x)| \le C(1+|x|), \quad x \in \mathbb{R}^d, \ t \in [0,T]$$
 (4.2)

$$|a(t,x) - a(t,y)| + |\sigma(t,x) - \sigma(t,y)| \le D|x - y|, \quad x, y \in \mathbb{R}^n, \ t \in [0,T]$$
 (4.3)

for some constants C and D and  $Z \in L^2(\Omega)$ . Then the SDE (4.1) has a unique continuous solution X(t) in the sense of Definition 4.1. Moreover, there exists K = K(C,T) such that

$$\mathbb{E}[|X(t)|^2] \le K(1 + \mathbb{E}[Z^2])e^{Kt}.$$

*Proof.* We first prove its uniqueness. Suppose there exist two solutions X and Y of (4.1) with continuous paths a.s. Then for all  $t \in [0, T]$ , we have

$$X(t) - Y(t) = \int_0^t (a(s, X(s)) - b(s, Y(s))) ds + \int_0^t (\sigma(s, X(s)) - \sigma(s, Y(s))) dB(s).$$

We now use Ito-isometry, Cauchy-Schwartz inequality, together with the assumption (4.3) to have

$$\begin{split} & \mathbb{E}\Big[|X(t)-Y(t)|^2\Big] \\ & \leq 2\mathbb{E}\Big[\Big(\int_0^t \left(a(s,X(s))-a(s,Y(s))\right)ds\Big)^2 + \Big(\int_0^t \left(\sigma(s,X(s))-\sigma(s,Y(s))\right)dB(s)\Big)^2\Big] \\ & \leq 2T\mathbb{E}\Big[\int_0^t \left(a(s,X(s))-a(s,Y(s))\right)^2ds\Big] + 2\mathbb{E}\Big[\int_0^t \left(\sigma(s,X(s))-\sigma(s,Y(s))\right)^2ds\Big] \\ & \leq 2D^2(T+1)\int_0^t \mathbb{E}\Big[|X(s)-Y(s)|^2\Big]ds \end{split}$$

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Hence by Grownwall's lemma,  $\mathbb{E}\Big[|(X(t)-Y(t)|^2\Big]=0$  for all  $t\in[0,T]$ . Hence

$$\mathbb{P}\Big(|X(t) - Y(t)| = 0 \text{ for all } t \in \mathbb{Q} \cap [0, T]\Big) = 1$$

where  $\mathbb Q$  denotes the rational numbers. By continuity  $t\mapsto |X(t)-Y(t)|$ , it follows that

$$\mathbb{P}\Big(X(t) = Y(t) \text{ for all } t \in [0, T]\Big) = 1.$$

This completes the uniqueness proof.