

# FE-610 Stochastic Calculus for Financial Engineers

Student : Riley Heiman  
Instructor: Thomas Lonon PhD

## Assignment #4

5.1, 5.2, 5.6, 5.7, 5.10

5.1

a.) Let  $f(x) = S(0) e^x$  and

$$X(t) = \int_0^t \sigma(s) dW(s) + \int_0^t \left( \alpha(s) - R(s) - \frac{1}{2} \sigma^2(s) \right) ds$$

Find  $df(X(t))$

$$f(t, X(t)) = S(0) e^{X(t)}$$

$$f(a, b) = S(0) e^b$$

$$f_a = 0$$

$$f_b = S(0) e^b = f_{bb}$$

$$df(t, X(t)) = S(0) e^{X(t)} dX(t) + \frac{1}{2} S(0) e^{X(t)} (dX(t))^2$$

$$df(t, X(t)) = \underbrace{S(0) e^{X(t)}}_{f(X(t))} \underbrace{\left[ dX(t) + \frac{1}{2} dX(t)^2 \right]}_{(*)}$$

What is  $dX(t) + (dX(t))^2$ ?

$$X(t) = \int_0^t \sigma(s) dW(s) + \int_0^t \left( \alpha(s) - R(s) - \frac{1}{2} \sigma^2(s) \right) ds$$

$$dX(t) = \sigma(t) dW(t) + \left( \alpha(t) - R(t) - \frac{1}{2} \sigma^2(t) \right) dt$$

$$(dX(t))^2 = (\underbrace{\sigma(t) dW(t) + \left( \alpha(t) - R(t) - \frac{1}{2} \sigma^2(t) \right) dt}_{\text{Ito's Lemma}})^2$$

$$= \sigma^2(t) dt$$

$$\oplus \left( dX(t) + \frac{1}{2} dX(t)^2 \right)$$

$$\sigma(t) dW(t) + \alpha(t) dt - R(t) dt - \frac{1}{2} \sigma^2(t) dt$$

$$(dX(t))^2 \longrightarrow \frac{+ \frac{1}{2} \sigma^2(t) dt}{\text{Cancels out}} \Rightarrow$$

$$\sigma(t) dW(t) + (\alpha(t) - R(t)) dt \quad (\text{put it together})$$

$$df(X(t)) = f(X(t)) \cdot [\sigma(t) dW(t) + (\alpha(t) - R(t)) dt]$$

5.1 (ii) Use Itô product Rule + Simplify

$$d(D(t)S(t)) = S(t)dD(t) + D(t)dS(t) + dD(t)dS(t)$$

$$dS(t) = \alpha(t)S(t)dt + \sigma(t)S(t)dW(t)$$

$$S(t) = S(0) \exp \left\{ \int_0^t \sigma(s) dW(s) + \int_0^t \left( \alpha(s) - \frac{1}{2} \sigma^2(s) \right) ds \right\}$$

$$dD(t) = -R(t)D(t)dt$$

$$D(t) = e^{-\int_0^t R(s)ds}$$

$$S(t)dD(t) + D(t)dS(t) + \underbrace{dD(t)dS(t)}_0 \Rightarrow$$

$$d(D(t)S(t)) = -S(t)R(t)D(t)dt + D(t)\alpha(t)S(t)dt + D(t)\sigma(t)S(t)dW(t)$$

$$d(D(t)S(t)) = S(t)D(t) \left[ (-R(t) + \alpha(t))dt + \sigma dW(t) \right]$$

5.2 Show eq. (5.2.30) can be written as

$$D(t)Z(t)V(t) = E[D(\tau)Z(\tau)V(\tau) | \mathcal{F}(t)]$$

$$(5.2.30) = D(t)V(t) = \tilde{E}[D(\tau)V(\tau) | \mathcal{F}(t)]$$

$$(5.2.11) = Z(t) = \exp \left\{ - \int_0^t \Theta(u) dW(u) - \frac{1}{2} \int_0^t \Theta^2(u) du \right\}$$

$$(5.2.21) = \Theta(t) = \frac{\alpha(t) - R(t)}{\sigma(t)}$$

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(5.2.30)

$$D(t)V(t) = \tilde{E}[D(\tau)V(\tau) | \mathcal{F}(t)] \Rightarrow$$

$$D(t)V(t) = \tilde{E}\left[D(\tau)V(\tau) \frac{Z(\tau)}{Z(t)} \mid \mathcal{F}(t)\right] \Rightarrow$$

$$D(t)V(t) = \frac{1}{Z(t)} \tilde{E}[D(\tau)V(\tau)Z(\tau) | \mathcal{F}(t)] \Rightarrow$$

$$D(t)V(t)Z(t) = \tilde{E}[D(\tau)V(\tau)Z(\tau) | \mathcal{F}(t)]$$

5.6 Use 2-D Levy (4.6.5)  
 to prove 2-D Girsanov Thm.  
 (5.4.1)

$$Z(t) = \exp \left\{ - \int_0^t \Theta(u) dW(u) - \frac{1}{2} \int_0^t \underbrace{\|\Theta\|^2}_{\text{Euclidean norm}} du \right\}$$

$$\sum_{j=1}^d \Theta_j^2(u)$$

Also,

$\Theta(t) = (\Theta_1(t), \dots, \Theta_d(t))$  is a  $d$  dim.  
 adapted process.

Thm. (5.2.1a)  $\tilde{W}(t) = W(t) + \int_0^t \Theta(u) du$

$$Z(t) = \exp \left\{ - \int_0^t \Theta(u) dW(u) - \frac{1}{2} \int_0^t \Theta^2(u) du \right\}$$

(\*)  $\left\{ \begin{array}{l} \text{Set } Z = Z(t). \text{ Show } W(t) \\ \text{is BM under } \tilde{\mathbb{P}}, \text{ then} \\ \tilde{W}(t) \text{ is also Brownian motion} \end{array} \right.$

## 5.6 cont.

use Lévy

①  $\tilde{W}(0) = 0$  ☒

$$\tilde{W}(t) = W(t) + \int_0^t \Theta(u) du$$

Set  $t=0$

$$\tilde{W}(0) = W(0) + \underbrace{\int_0^0 \Theta(u) du}_0 = 0$$

②  $\tilde{W}(t)$  depends on  $W(t)$  which is cont., so  $\tilde{W}(t)$  must also be cont.

③  $[\tilde{W}_i, \tilde{W}_i](t) = \int_0^t (dW_i(u))^2$

$$\tilde{W}(t) = W(t) + \int_0^t \Theta(u) du \Rightarrow$$

$d\tilde{W}(t) = dW(t) + \Theta(t) dt$  (what is  $d\tilde{W}(t)^2$ ?)

$$(dW(t) + \Theta(t) dt)(dW(t) + \Theta(t) dt)$$

$$\underbrace{\hspace{10em}}_{\underline{dt}}$$

$$\int_0^t du = t - 0 = t$$

The  $QV = t$  ☒

(4)  $\tilde{W}_1$  is independent from  $\tilde{W}_2$   
if  $[\tilde{W}_1, \tilde{W}_2](t) = 0$

$$d\tilde{W}_1(t) d\tilde{W}_2(t) = 0$$

(5) Show  $\tilde{W}(t)$  is a martingale.

Lemma 5.2.2

$$\tilde{E}[Y | \mathcal{F}(s)] = \frac{1}{Z(s)} E[Y Z(t) | \mathcal{F}(s)]$$

$$\begin{aligned} (*) \quad \tilde{E}[\tilde{W}(t) | \mathcal{F}(s)] &= \frac{1}{Z(s)} E[\tilde{W}(t) Z(t) | \mathcal{F}(s)] \\ &= \tilde{W}(s) \end{aligned}$$

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$$\frac{1}{Z(s)} \tilde{E}[\tilde{W}(t) Z(t) | \mathcal{F}(s)] = \frac{1}{Z(s)} \cdot \tilde{E}[\tilde{W}(t) | \mathcal{F}(s)] \cdot \underbrace{E[Z(t) | \mathcal{F}(s)]}_{Z(s)}$$

$$\Rightarrow \frac{Z(s)}{Z(s)} \cdot \tilde{E}[\tilde{W}(t) | \mathcal{F}(s)] = \tilde{W}(s)$$

By the Girsanov Thm.  $\tilde{W}(t)$  is  
a Brownian Motion.



## 5.7 (i)

(arbitrage Thm. 5.4.23)

$$\mathbb{P}(X(\tau) \geq 0) = 1, \quad \mathbb{P}(X(\tau) > 0) > 0$$

$$+ X(0) = 0$$

Show if  $X_2(0)$  is positive ( $X_2(0) > 0$ )  
then

$$\mathbb{P}\left\{X_2(\tau) \geq \frac{X_2(0)}{D(\tau)}\right\} = 1, \quad \mathbb{P}\left\{X_2(\tau) > \frac{X_2(0)}{D(\tau)}\right\} > 0$$

$$\text{Let } D(\tau) = e^{-r(\tau-t)}$$

$$\mathbb{P}\left\{X_2(\tau) \geq \underbrace{X_2(0)}_{\text{This term is positive}} \cdot \underbrace{e^{+r(\tau-t)}}_{\text{This term is positive + constant.}}\right\}$$

This term  
is  
positive

This term is  
positive + constant.

$$\text{(i.e.) } \$5 \cdot e^{0.02(1 \text{ year})} = \$5.10$$

$$X_2(\tau) \text{ is } \geq X_2(0) e^{r(\tau-t)} \quad \text{so } \mathbb{P} = 1$$

## 5.7 (ii)

Show that if a multidimensional Market Model has a portfolio Value process  $X_2(\tau)$  s.t.  $X_2(0)$  is positive and (5.4.24) holds, then the model has portfolio Value process  $X_1(0)$  s.t.  $X_1(0) = 0$  and (5.4.23) holds.

(5.4.24)

$$\mathbb{P}\left\{X(\tau) \geq \frac{X(0)}{D(\tau)}\right\} = 1, \quad \mathbb{P}\left\{X(\tau) > \frac{X(0)}{D(\tau)}\right\} > 0$$

(5.4.23)

$$\mathbb{P}(X(\tau) \geq 0) = 1, \quad \mathbb{P}(X(\tau) > 0) > 0$$

If Thm. (5.4.24) holds, then

$$\mathbb{P}\left\{X(\tau) \geq \frac{X(0)}{D(\tau)}\right\} = 1$$

Since  $X_2(0)$  is positive then  $\frac{X_2(0)}{D(t)} > 0$

So  $\mathbb{P}\{X(\tau) \geq 0\} = 1$ , which is exactly

Thm. 5.2.23

## 5.10 (Chooser option)

(i) Show @  $t_0$  the value of Chooser option is:

$$C(t_0) + (-F(t_0))^+$$

Let the Value =  $\max \{C_0, P_0\}$ ,

$$\text{option Value} = \max \{C_0, \underbrace{C_0 - F_0}_{C_t - P = F, \text{ so } P = C - F}\}$$

$$\text{Option Value} = C(t_0) + (-F_0)^+$$

(ii) Show the value @  $t_0 \Rightarrow$

$$\text{Value}_{t_0} = C(T) + P(t_0) \quad \uparrow \quad K = e^{-r(T-t_0)} K$$

If  $T = t_0$ , then

$$K = e^{-r(\text{red } t_0 - t_0)} K = e^0 K = K$$

$$\Rightarrow \text{Value}(t_0) = C(T) + P(t_0)$$