#### 1 Proof properties of Wiener process

Assume that t < s

$$X_s \sim N(0,s)$$
  $X_t \sim N(0,t)$ 

- 1. By independent increments, we know that  $E(X_s|X_t=B)=E(X_s-X_t+X_t|X_t=B)=E(X_s-X_t|X_t=B)+X_t=0+B=B$
- 2.  $Var(X_s|X_t = B) = E\{[X_s E(X_s|X_t = B)]^2 | X_t = B\} = E[(X_s B)^2 | X_t = B]$   $= E[X_s^2 - 2X_sB + B^2 | X_t = B] = E[X_s^2 | X_t = B] - 2BE(X_s | X_t = B) + B^2$  $= E[X_s^2 | X_t = B] - B^2 = Var(X_s) - Var(X_t) = s - t$
- 3.  $X_t|X_s=B$  is a bivariate normal distribution which  $\sim N(B\times \frac{t}{s},(s-t)\times \frac{t}{s})$ . So  $E(X_t|X_s=B)=B\times \frac{t}{s}$
- 4. According to 3.  $Var(X_t|X_s=B)=(s-t)\times \frac{t}{s}$
- 5.  $Cov(X_s, X_t) = E(X_s X_t) E(X_s)E(X_t) = E(X_s X_t) 0 = E(X_s X_t)$   $= E(X_s - X_{s-1} + X_{s-1} - X_{s-2} + \dots + X_{t+1} - X_t + X_t)X_t$  $= E(X_s - X_{s-1})X_t + \dots + E(X_t)X_t = E(X_t^2) = t^2$

#### 2 Proof the Quadratic Variation of Brownian Motion

According to standard Brownian Motion

$$dX_t = adt + bdW_t$$

By definition,  $(dZ)^2 = \varepsilon^2 dt$ ,  $\forall \varepsilon \sim N(0,1)$ . So

$$Var(\varepsilon) = 1 = E[\varepsilon^2] \Rightarrow E[(dZ)^2] = E[\varepsilon^2 dt] = 1dt = dt$$

In addition,  $Var((dZ^2)) = Var(\varepsilon^2 dt) = (dt^2)Var(\varepsilon) \to 0$  (because  $(dt^2 \to 0)$ ). So we say  $(dZ)^2 \stackrel{a.s.}{=} dt$ 

Based on the result of  $(dZ)^2 = dt$ , we can infer that the quadratic variation of Brownian Motion over [0, T], i.e.,  $[W, W](t) = \lim_{n \to \infty} \sum_{i=1}^{n} |B(t_i^n) - B(t_{i-1}^n)|^2 = T$ .

<Proof>

At first we know  $(dZ)^2 \sim N(dt,0)$  when  $n \to \infty$  or  $dt \to 0$ ,  $(B(t_i^n) - B(t_{i-1}^n)) \stackrel{a.s.}{=} (t_i^n - t_{i-1}^n)$ . This is because  $E[(B(t_i^n) - B(t_{i-1}^n))^2] = E[\varepsilon^2(t_i^n - t_{i-1}^n)] = t_i^n - t_{i-1}^n$ , and  $Var((B(t_i^n) - B(t_{i-1}^n))^2) = Var(\varepsilon^2(t_i^n - t_{i-1}^n)) = (t_i^n - t_{i-1}^n)^2 Var(\varepsilon^2) \to 0$ .

Thus, we can conclude that when  $n \to \infty$ , or  $t_i^n - t_{i-1}^n \to 0$ ,  $\lim_{n \to \infty} \sum_{i=1}^n |B(t_i^n) - B(t_{i-1}^n)|^2 = T$ 

# 3 Application of Itô's Lemma (I)

Assume the stock price change process is

$$dS_t = \mu S_t dt + \sigma S_t dz_t$$
$$f_t = \ln S_t$$

According to Taylor series

$$df_t = \left(\frac{\partial f}{\partial t}dt + \frac{\partial f}{\partial S}\mu S_t(S_t, t) + \frac{1}{2}\frac{\partial^2 f}{\partial S^2}\sigma^2 S_t^2(S_t, t)\right)dt + \frac{\partial f}{\partial S}\sigma S_t(S_t, t)dz_t$$

Because  $f_t = \ln S_t$ 

$$d \ln S_t = \left(\frac{1}{S_t} \mu S_t + \frac{-1}{2S_t^2} \sigma^2 S_t^2\right) dt + \frac{1}{S_t} \sigma S_t dz_t$$
$$= \left(\mu - \frac{1}{2} \sigma^2\right) dt + \sigma dz_t$$

Then we inergrate it to get the close form of stock price.

$$\int_{0}^{T} d \ln S_{t} = \int_{0}^{T} (\mu - \frac{1}{2}\sigma^{2}) dt + \int_{0}^{T} \sigma dz_{t}$$
$$\ln(S_{T} - S_{0}) = (\mu - \frac{1}{2}\sigma^{2})T + \sigma z_{T}$$
$$(S_{T} - S_{0}) = \exp\left[(\mu - \frac{1}{2}\sigma^{2})T + \sigma z_{T}\right]$$
$$S_{T} = S_{0} + \exp\left[(\mu - \frac{1}{2}\sigma^{2})T + \sigma z_{T}\right]$$

Expectation value and variance

## 4 Application of Itô's Lemma (II)

Assume the stock price change process is

$$dS_t = \mu S_t dt + \sigma S_t dz_t$$

$$f_t = S_t - Ke^{-r(T-t)}$$

According to Taylor series

$$df_t = (-rK^{-r(T-t)} + \mu S_t + \frac{(\sigma S_t)^2}{2} \times 0)dt + \sigma S_t dz_t$$
$$= (\mu S_t - rK^{-r(T-t)})dt + \sigma S_t dz_t$$

## 5 Application of Itô's Lemma (III)

Assume the interest rate change process is

$$dr_t = \kappa(\theta - r_t)dt + \sigma dz_t$$
$$f_t = r_t e^{\kappa t}$$

According to Taylor series

$$df_{t} = \left(\frac{\partial f}{\partial t} + \frac{\partial f}{\partial r}\kappa(\theta - r_{t}) + \frac{1}{2}\frac{\partial^{2} f}{\partial r^{2}}\sigma^{2}(r_{t}, t)\right)dt + \frac{\partial f}{\partial r}\sigma(r_{t}, t)dz_{t}$$
$$= \left(\kappa r_{t}e^{\kappa t} + e^{\kappa t}\kappa(\sigma - r_{t})\right)dt + e^{\kappa t}\sigma dz_{t}$$

Because  $f_t = r_t e^{\kappa t}$ 

$$dr_t e^{\kappa t} = (\kappa r_t e^{\kappa t} + e^{\kappa t} \kappa (\sigma - r_t)) dt + e^{\kappa t} \sigma dz_t$$
$$dr_t = (\kappa r_t + \kappa (\sigma - r_t)) dt + \sigma dz_t$$
$$= \kappa \sigma dt + \sigma dz_t$$

Then we inergrate it to get the close form of interest rate.

$$\int_0^T dr_t = \int_0^T \kappa \sigma dt + \int_0^T \sigma dz_t$$
$$r_T - r_0 = \int_0^T \kappa \sigma dt + \int_0^T \sigma dz_t$$
$$r_T = r_0 + \kappa \sigma T + \sigma z_T$$

Exepcted Value and Variance

## 6 Application of Itô's Lemma (IV)

Assume the stock price change process and foriegn exchange rate change process are

$$dS_t = \mu S_t dt + \sigma S_t dz_{St}$$
  
$$dX_t = \mu X_t dt + \sigma X_t dz_{Xt}$$
  
$$Cov(dz_{St}, dz_{Xt}) = \rho dt$$

if  $f_t = f(t, X_t, S_t) = S_t X_t$ , According to Taylor series

$$df_{t} = \left(\frac{\partial f_{t}}{\partial t} + \frac{\partial f_{t}}{\partial S_{t}}\mu_{S}S_{t} + \frac{\partial f_{t}}{\partial X_{t}}\mu_{X}X_{t} + \frac{1}{2}\frac{\partial^{2} f_{t}}{\partial S_{t}^{2}}\sigma_{S}^{2}S_{t}^{2} + \frac{1}{2}\frac{\partial^{2} f_{t}}{\partial X_{t}^{2}}\sigma_{X}^{2}X_{t}^{2} + \frac{\partial^{2} f_{t}}{\partial S_{t}\partial X_{t}}\sigma_{S}S_{t}\sigma_{X}X_{t}\rho\right)dt$$

$$+ \frac{\partial f_{t}}{\partial S_{t}}\sigma_{S}S_{t}dz_{St} + \frac{\partial f_{t}}{\partial X_{t}}\sigma_{X}X_{t}dz_{Xt}$$

$$= \left(X_{t}\mu_{S}S_{t} + S_{t}\mu_{X}X_{t} + \sigma_{S}\sigma_{X}S_{t}X_{t}\rho\right)dt + X_{t}\sigma_{S}S_{t}dz_{St} + S_{t}\sigma_{X}X_{t}dz_{Xt}$$

Because  $f_t = S_t X_t$ 

$$dS_t X_t = (X_t \mu_S S_t + S_t \mu_X X_t + \sigma_S \sigma_X S_t X_t \rho) dt + X_t \sigma_S S_t dz_{St} + S_t \sigma_X X_t dz_{Xt}$$

Then we inergrate it to get the close form of stock price with NTD dollars.

$$\int_{o}^{T} dS_{t}X_{t} = \int_{0}^{T} (X_{t}\mu_{S}S_{t} + S_{t}\mu_{X}X_{t} + \sigma_{S}\sigma_{X}S_{t}X_{t}\rho)dt + \int_{o}^{T} X_{t}\sigma_{S}S_{t}dz_{St} + \int_{o}^{T} S_{t}\sigma_{X}X_{t}dz_{Xt}$$

$$(S_{T} - S_{0})(X_{T} - X_{0}) = \int_{0}^{T} (S_{t}X_{t}(\mu_{S} + \mu_{X} + \sigma_{S}\sigma_{X}\rho))dt + \int_{0}^{T} S_{t}X_{t}(\sigma_{S}dz_{St} + \sigma_{X}dz_{Xt})$$

$$= [(S_{T}X_{T} - S_{0}X_{0})(\mu_{S} + \mu_{X} + \sigma_{S}\sigma_{X}\rho)]T + (S_{T}X_{T} - S_{0}X_{0})(\sigma_{S}dz_{ST} + \sigma_{X}dz_{XT})$$