

$$d(AB) = AdB + BdA + dAdB$$

Feynman Kac:

$$\frac{\partial F}{\partial t} + \mu(t, x) \frac{\partial F}{\partial x} + \frac{1}{2} \sigma^2(t, x) \frac{\partial^2 F}{\partial x^2} - r(x)F = -h(x)$$

$$F(T, x) = \Phi(x)$$

$$dX = \mu(t, X)dt + \sigma(t, X)dW$$

$$F(t, x) = E_t \left[\int_t^T e^{-\int_t^s r(X_u)du} h(X_s)ds + e^{-\int_t^T r(X_u)du} \Phi(X_T) \right]$$

Geometric Brownian Motion:

$$dX_t = \alpha X_t dt + \sigma X_t dW_t \quad X \text{ is log normal}$$

$$\log X_T \sim N \left\{ \log X_t + \left(\alpha - \frac{1}{2} \sigma^2 \right) (T-t); \sigma^2 (T-t) \right\}$$

if $T = t, t = 0:$

$$\log X_t \sim N \left\{ \log X_0 + \left(\alpha - \frac{1}{2} \sigma^2 \right) t; \sigma^2 t \right\}$$

$$E_t(X_T) = X_t e^{\alpha(T-t)}$$

$$Var(X_T) = X_T^2 [e^{(2\alpha-\sigma^2)(T-t)} - e^{2\alpha(T-t)}]$$

$$E_0(X_t) = X_0 e^{\alpha t}$$

$$Var(X_t) = X_0 [e^{(2\alpha-\sigma^2)t} - e^{2\alpha t}]$$

For a log normal distribution:

$$X \sim N\{\mu; \sigma^2\} \quad E(X) = e^{\mu + \frac{1}{2}\sigma^2}$$

$$Var(X) = e^{2(\mu+\sigma^2)} - e^{2\mu+\sigma^2} \quad E(X^2) = e^{2\mu+2\sigma^2}$$

$$E(X^n) = E^{n\mu + \frac{1}{2}n^2\sigma^2}$$

$$Var(X) = E(X^2) - (E(X))^2$$

Moment generating function of normal:

$$M_x(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2} \quad E(X) = M'_x(0) = \mu$$

$$E(X^2) = M''_x(0) = \sigma^2 + \mu \quad E(X^n) = M^{(n)}_x(0)$$

Risk free asset is that it has no diffusion dW term:

$$dB(t) = rB(t)dt$$

$$dS(t) = \alpha S(t)dt + \sigma S(t)d\bar{W}(t)$$

No Arbitrage approach to Black Schole

$$\Pi(t) = F(t, S)$$

$$d\Pi = \alpha_\pi \Pi dt + \sigma_\pi \Pi d\bar{W}(t)$$

$$\alpha_\pi(t) = \frac{F_t + \alpha S F_s + \frac{1}{2} \sigma^2 S^2 F_{ss}}{F}$$

$$\sigma_\pi(t) = \frac{\sigma S F_s}{F}$$

Consider a portfolio based on: 1. The stock,

2. The derivative asset:

$$dV = V[u_s \alpha + u_\pi \alpha_\pi]dt + V[u_s \sigma + u_\pi \sigma_\pi]d\bar{W}$$

$$u_s + u_\pi = 1 \quad u_s \alpha + u_\pi \alpha_\pi = r \quad u_s \sigma + u_\pi \sigma_\pi$$

$$u_s = \frac{\sigma_\pi}{\sigma_\pi - \sigma} \quad u_\pi = \frac{-\sigma}{\sigma_\pi - \sigma}$$

$$u_s = \frac{SF_s}{SF_s - F} \quad u_\pi = \frac{-F}{SF_s - F}$$

Black Scholes Equation:

$$F_t + rSF_s + \frac{1}{2} S^2 \sigma^2 F_{ss} - rF = 0$$

$$F(T, s) = \Phi(s)$$

Under Q-Measurement, Risk Neutral Valuation:

$$dS = rSdt + S\sigma dW$$

$$F(t, s) = e^{-r(T-t)} E_{t,s}^Q [\Phi(S(T))]$$

By Ito's lemma:

$$S(T) = s \cdot e^{(r - \frac{1}{2}\sigma^2)(T-t) + \sigma(W(T) - W(t))}$$

$$z = \left(r - \frac{1}{2}\sigma^2 \right) (T-t) + \sigma(W(T) - W(t))$$

$$z \sim N \left\{ \left(r - \frac{1}{2}\sigma^2 \right) (T-t); \sigma^2 (T-t) \right\}$$

$$F(t, s) = e^{-r(T-t)} \int_{-\infty}^{\infty} \Phi(se^z) f(z) dz$$

$$E(g(x)) = \int_{-\infty}^{\infty} g(x) f(x) dx \quad f(x) \text{ is the pdf of } x$$

$$\text{pdf of normal: } \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$E^Q[\max(se^z - K, 0)] = 0 \cdot Q(se^z < K) + \int_{\log(\frac{K}{s})}^{\infty} (se^z - K) f(z) dz$$

$$F(t, s) = sN(d_1(t, s)) - e^{-r(T-t)} KN(d_2(t, s))$$

$$d_1(t, s) = \frac{1}{\sigma\sqrt{T-t}} \left\{ \log\left(\frac{s}{K}\right) + \left(r + \frac{1}{2}\sigma^2 \right) (T-t) \right\}$$

$$d_2(t, s) = d_1(t, s) - \sigma\sqrt{T-t}$$

$$\text{For a forward, } \Phi(S_T) = S_T - K$$

$$f(t) = e^{r(T-t)} S_t \quad K = S_0 e^{rT}$$

Replication Method to Black Sholes:

$$u^0 + u^* = 1 \quad u^0: \text{Bond weight, } u^*: \text{Stock weight}$$

$$dV = V\{u^0 r + u^* \alpha\}dt + V u^* \sigma d\bar{W}(t)$$

$$V(T) = \Phi(S(T))$$

$V(t) = F(t, S)$ By Ito's Lemma:

$$dV = \left\{ F_t + \alpha S F_s + \frac{1}{2} \sigma^2 S^2 F_{ss} \right\} dt + \sigma S F_s d\bar{W}$$

$$dV = V \left\{ \frac{F_t + \alpha S F_s + \frac{1}{2} \sigma^2 S^2 F_{ss}}{F} \right\} dt + V \frac{S F_s}{F} \sigma d\bar{W}$$

$$u^* = \frac{S F_s}{F} \quad u^0 = \frac{F_t + \frac{1}{2} \sigma^2 S^2 F_{ss}}{r F} = 1 - \frac{S F_s}{F}$$

$$h^0 = \frac{u^0 V}{B} = \frac{F - S F_s}{B} \quad h^* = \frac{u^* V}{S} = F_s$$

The Greeks:

$$\Delta = \frac{\partial P}{\partial S} = N(d_1)$$

$$\Gamma = \frac{\partial^2 P}{\partial S^2} = \frac{\varphi(d_1)}{s\sigma\sqrt{T-t}} \quad \varphi(d_1) \text{ is the pdf of } N(0, 1)$$

$$\rho = \frac{\partial P}{\partial r} = K(T-t)e^{-r(T-t)} N(d_2)$$

$$\Theta = \frac{\partial P}{\partial t} = -\frac{s\varphi(d_1)\sigma}{2\sqrt{T-t}} - rKe^{-r(T-t)} N(d_2)$$

$$v = \frac{\partial P}{\partial \sigma} = s\varphi(d_1)\sqrt{T-t}$$

Put-Call Parity:

$$P(t, s) + s = Ke^{-r(T-t)} + c(t, s)$$

Multidimensional Black Sholes

$$dS_i = \alpha_i S_i dt + S_i \sum_{j=1}^n \sigma_{ij} d\bar{W}_i(t)$$

σ is $n \times n$ matrix $\{\sigma_{ij}\}$

$$dF = F \cdot \alpha_F dt + F \cdot \sigma_F d\bar{W}$$

If 2 dimension:

$$dF = \left[\frac{\partial F}{\partial t} + \frac{\partial F}{\partial S_1} \alpha_1 S_1 + \frac{\partial F}{\partial S_2} \alpha_2 S_2 \right.$$

$$+ \frac{1}{2} \frac{\partial^2 F}{\partial S_1^2} S_1^2 (\sigma_{11}^2 + \sigma_{12}^2) + \frac{1}{2} \frac{\partial^2 F}{\partial S_2^2} S_2^2 (\sigma_{21}^2 + \sigma_{22}^2)$$

$$+ \frac{\partial^2 F}{\partial S_1 \partial S_2} S_1 S_2 (\sigma_{11}\sigma_{21} + \sigma_{12}\sigma_{22}) \Big] dt$$

$$+ \left[\frac{\partial F}{\partial S_1} S_1 \sigma_1 + \frac{\partial F}{\partial S_2} S_2 \sigma_2 \right] dW$$

$$\alpha_F = \frac{1}{F} \left[F_t + \sum_1^n \alpha_i S_i F_i + \frac{1}{2} \text{tr}\{\sigma^T D[S] F_{ss} D[S] \sigma\} \right]$$

$$\sigma_F = \frac{1}{F} \sum_1^n S_i F_i \sigma_i$$

$$F_{ss} = \left\{ \frac{\partial^2 F}{\partial S_i \partial S_j} \right\}_{i,j=1}^n$$

$$u_B = 1 - \left(\sum_1^n u_i + u_F \right)$$

$$dV = V \left[\sum_1^n u_i \frac{dS_i}{S_i} + u_F \frac{dF}{F} + u_B \frac{dB}{B} \right]$$

$$dV = V \left[\sum_1^n u_i (\alpha_i - r) + u_F (\alpha_F - r) + r \right] dt$$

$$+ V \left[\sum_1^n u_i \sigma_i + u_F \sigma_F \right] d\bar{W}$$

$$\sum_1^n u_i \sigma_i + u_F \sigma_F = 0$$

$$\begin{bmatrix} \alpha_1 - r & \dots & \alpha_n - r & \alpha_F - r \\ \sigma_1^T & \dots & \sigma_n^T & \sigma_F^T \end{bmatrix} \begin{bmatrix} u_S \\ u_F \end{bmatrix} = \begin{bmatrix} \beta \\ 0 \end{bmatrix}$$

σ_i^T is column vector

$$u_S = \begin{bmatrix} u_1 \\ u_2 \\ \dots \\ u_n \end{bmatrix} \quad H =$$

$$\begin{bmatrix} \alpha_1 - r & \dots & \alpha_n - r & \alpha_F - r \\ \sigma_1^T & \dots & \sigma_n^T & \sigma_F^T \end{bmatrix}$$

H must be singular

$$\alpha_i - r = \sum_{j=1}^n \sigma_{ij} \lambda_j, \quad i = 1, \dots, n,$$

$$\alpha_F - r = \sum_{j=1}^n \sigma_{Fj} \lambda_j, \quad \lambda = \begin{bmatrix} \lambda_1 \\ \dots \\ \lambda_n \end{bmatrix}$$

$$\alpha - r \mathbf{1}_n = \sigma \lambda \quad \alpha_F - r = \sigma_F \lambda$$

$$F_t + \sum_{i=1}^n r S_i F_i + \frac{1}{2} \text{tr}\{\sigma^T D[S] F_{ss} D[S] \sigma\} - rF = 0$$

$$F = e^{-r(T-t)} E^Q[\Phi(S(T))]$$

Reducing the State Space

$$F(t, s_1, \dots, s_n) = s_n G \left(t, \frac{s_1}{s_n}, \dots, \frac{s_{n-1}}{s_n} \right)$$

$$z = \left(\frac{s_1}{s_n}, \dots, \frac{s_{n-1}}{s_n} \right)$$

$$F_t(t, s) = s_n G_t(t, z)$$

$$F_i(t, s) = G_i(t, z), i = 1, \dots, n-1$$

$$F_n(t, s) = G(t, z) - \sum_{j=1}^{n-1} \frac{s_j}{s_n} G_j(t, z)$$

$$F_{ij}(t, s) = \frac{1}{s_n} G_{ij}(t, z), \quad i, j = 1, \dots, n-1$$

$$F_{in}(t, s) = F_{ni}(t, s) = - \sum_{j=1}^{n-1} \frac{s_j}{s_n^2} G_{ij}(t, z),$$

$i = 1, \dots, n-1$

$$F_{nn} = - \sum_{i,j=1}^{n-1} \frac{s_i s_j}{s_n^3} G_{ij}(t, z)$$

For two dimension:

$$F_1 = G_z \quad F_2 = G - zG_z$$

$$F_{11} = \frac{1}{S_2} G_{zz} \quad F_{22} = \frac{S_1^2}{S_2^3} G_{zz}$$

$$G_t + \frac{1}{2}(\sigma_1^2 + \sigma_2^2)z^2 G_{zz} = 0$$

$$F(t, s_1, s_2) = s_1 N(d_1(t, z)) - s_2 N(d_2(t, z))$$

$$d_1(t, z) = -\frac{1}{\sqrt{(\sigma_1^2 + \sigma_2^2)(T-t)}} \left\{ \log z + \frac{1}{2}(\sigma_1^2 + \sigma_2^2)(T-t) \right\}$$

$$d_2(t, z) = d_1(t, z) - \sqrt{(\sigma_1^2 + \sigma_2^2)(T-t)}$$

Dividends

$\delta = \delta(S_{t-})$ is a function of S_{t-}

$$S_t = S_{t-} - \delta(S_{t-})$$

$$F^0(T, S_T) = \Phi(S_T)$$

$$F^1(T_1^-, S_{T^-}) = F^0(T_1, S_{T^-} - \delta(S_{T^-}))$$

$$F(t, S_t) = e^{-r(T-t)} E^Q[\Phi(S_T)]$$

$$dS_t = rS_t dt + \sigma S_t dW_t$$

if $\delta(S_{t-}) = \delta S_{t-}$, δ is a constant:

$$F_\delta(t, S_t) = F(t, (1-\delta)^n S_t)$$

n is the number of dividend points in the interval $(t, T]$

Continuous dividends:

Q-measure:

$$dS = (r - \delta)Sdt + \sigma SdW$$

$$F(t, S_t) = e^{-r(T-t)} E^Q[\Phi(S_T)]$$

$$F_\delta(t, s) = F_0(t, se^{-\delta(T-t)})$$

Forward with dividend:

$$K = S_0 e^{(r-\delta)T}$$

Foreign Exchange:

P measure:

$$dX_t = X_t \alpha dt + X_t \sigma_X d\bar{W}_t$$

Q measure:

$$dX_t = X_t (r^d - r^f) dt + X_t \sigma_X dW_t$$

This is similar to assuming that X_t is dividend paying asset.

$$\text{Forward: } K = X_0 e^{(r^d - r^f)T}$$

$$F_\delta(t, S_t) = F(t, e^{-\delta(T-t)} S_t)$$

$$F(t, x) = x e^{-r^f(T-t)} N[d_1] - e^{-r^d(T-t)} K N[d_2]$$

$$d_1(t, x) = \frac{1}{\sigma_X \sqrt{T-t}} \left\{ \log \left(\frac{x}{K} \right) + \left(r^d - r^f + \frac{1}{2} \sigma_X^2 \right) (T-t) \right\}$$

$$d_2(t, x) = d_1 - \sigma_X \sqrt{T-t}$$

$$\textcircled{5} \quad X_t = \int_0^t g(s) dW \sim N(0, \int_0^t g(s)^2 ds)$$

$$Z_t = e^{-\frac{1}{2}\eta^2 \int_0^t g^2(s) ds + \eta \int_0^t g(s) dW_s}$$

$$\textcircled{6} \quad \beta_k(t) = E[W_t^k] \quad \beta_{k+1}(t) = \frac{1}{2} k(k-1) \int_0^t \beta_{k+2}(s) ds$$

$$\textcircled{7} \quad dX_t = -\eta (X_t - \bar{X}) dt + \sigma dW_t \quad \text{moment-generating function} \\ f(t, x) = e^{\alpha_0 t + \alpha_1 t \lambda} \Rightarrow \begin{cases} \alpha_0(t) = \lambda e^{-\eta t} \\ \alpha_1(t) = \lambda \bar{X} (1 - e^{-\eta t}) + \frac{\lambda^2 t^2}{2\eta} (1 - e^{-2\eta t}) \end{cases}$$

$$\textcircled{8} \quad d\bar{\pi} = r \bar{\pi} (t, S_t) dt + \sigma \bar{\pi} (t, S_t) dW_t \quad \left. \begin{array}{l} \text{under Q} \\ \bar{\pi}_t = \frac{\frac{\partial}{\partial S} \pi(t, S_t)}{\pi(t, S_t)} \end{array} \right\}$$

$$\textcircled{9} \quad \Phi(S_T) = S_T \quad F(t, S_t) = S_t$$

$$\textcircled{10} \quad \Phi(S_t) = \frac{1}{S_t} \quad Z = \frac{1}{S_t} \quad F(t, S_t) = \frac{1}{S_t} e^{(2r + \sigma^2)(T-t)}$$

Binomial Pricing

$$\text{Arbitrage: } \text{PIV}(T; h) \geq 0 \quad \text{PIV}(T; h) > 0 > 0$$

Free of arbitrage for binomial pricing:

$$d \leq \frac{1+r}{1+r} \leq u$$

$$\text{Under } Q: \quad q_u = \frac{u-d}{u-d} \\ q_d = \frac{u-(1+r)}{u-d}$$

replicating portfolio:

market is complete = all claims can be replicated

$X = \Phi(S_T)$ is reachable: can be replicated

In binomial: $u > d$ = completeness

$$V_0^h = \frac{1}{1+r} E^Q[X] = \frac{1}{1+r} \{q_u \Phi(u) + q_d \Phi(d)\}$$

multi-period

$$\bar{\pi}(t; X) = V_0 = \frac{1}{(1+r)^T} E^Q[\Phi(S_T)]$$

$$\bar{\pi}(t; X) = \frac{1}{(1+r)^T} \sum_{k=0}^T \binom{T}{k} q_u^k q_d^{T-k} \Phi(S_0 u^k d^{T-k})$$

Stochastic Integrals

Brownian motion:

$$W(0)=0$$

$$W(u) - W(t), \quad W(s) - W(r) \text{ independent} \quad (r < s < t < u)$$

$$W(t) - W(s) \sim N(0, t-s) \quad (s < t)$$

W has continuous trajectories

Properties:

$$E \int_a^b g(s) dW(s) = 0$$

$$E[(\int_a^b g(s) dW(s))^2] = E \int_a^b g^2(s) ds$$

Martingales:

$$\left\{ \begin{array}{l} X \text{ adapted to } F_t \\ E[X_t | F_s] \text{ const} \\ E[X_t | F_s] = X(s) \quad (t > s) \end{array} \right.$$

martingale = no "dt" term

Itô's Lemma:

$$\int_0^t (dW)^2 = t \quad \Leftrightarrow \quad (dW)^2 = dt$$

$$dX = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dx + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} (dx)^2$$

$$dt^2 = 0 \quad dt dW = 0 \quad dW^2 = dt$$

Geometric Brownian Motion

$$dX_t = \alpha X_t dt + \sigma X_t dW_t \quad Z = \log(X_t) \\ X_T = e^{(\alpha - \frac{1}{2}\sigma^2)(T-t) + \sigma(W_T - W_t)} \quad E(X_T) = e^{\alpha(T-t)}$$

multi-dimensional Itô's formula

$$dX_i(t) = \mu_i(t) dt + \sum_{j=1}^d \sigma_{ij}(t) dW_j(t)$$

$$d^2f(t, X_t) = \frac{\partial^2 f}{\partial t^2} dt + \sum \frac{\partial^2 f}{\partial x_i \partial x_j} dx_i dx_j + \frac{1}{2} \sum \sum \frac{\partial^2 f}{\partial x_i^2} dX_i dX_j$$

$$dt^2 = 0 \quad dt dW = 0 \quad d^2W_i = dt \quad dW_i dW_j = 0 \quad (i \neq j)$$

product rule: $d(AB) = A \cdot dB + B \cdot dA + dA \cdot dB$

Stochastic Differential Equations

linear SDE $dX_t = \alpha X_t dt + \sigma dW_t$

$$X_t = X_0 e^{\alpha t} + \sigma \int_0^t e^{\alpha(t-s)} dW_s \quad (\text{normal distribution})$$

Feynman-Kac Theorem

$$\left\{ \begin{array}{l} \frac{\partial F(t, x)}{\partial t} + \frac{\partial F}{\partial x} \mu(t, x) F + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} \sigma^2(t, x) - r(x) F(t, x) = -h(x) \\ F(T, x) = \Phi(x) \\ \Rightarrow dX = \mu(t, x) dt + \sigma(t, x) dW_t \\ F(t, x) = E_t \left[\int_t^T e^{-\int_s^T r(x_u) du} h(X_s) ds + e^{-\int_t^T r(x_u) du} \Phi(X_T) \right] \end{array} \right.$$

moment-generating functions

$$\Phi(x_t) = e^{\lambda X_t} \quad f(t, x) = E[\Phi(X_t) | X_t = x]$$

$$E[d\Phi(t)] = 0 \Rightarrow \text{martingale}$$

Portfolio Dynamics

Self-financing:

$$\int V_t^h = \sum h_i(t) S_i(t)$$

$$dV_t^h = \sum h_i(t) dS_i(t)$$

For self-financing no arbitrage:

$$dV_t^h = k(t) V_t^h dt \quad (\Rightarrow k(t) = r(t))$$

Black-Scholes Assumptions (complete)

the derivative security can be bought and sold in market
free of arbitrage

B-S equation

$$\frac{\partial F}{\partial t} + r S_t \frac{\partial F}{\partial S} + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 F}{\partial S^2} - r F = 0$$

Boundary condition $F(T, S_T) = \Phi(S_T)$

solving: $dS_t = r S_t dt + \sigma S_t dW_t$ (非真實)

$$F(t, s) = e^{-r(T-t)} E_t^Q [\Phi(S_T) | S_t = s]$$

Call option

$$F(t, S_t) = S_t N(d_1) - e^{-r(T-t)} K N(d_2)$$

$$d_1 = \frac{1}{\sigma \sqrt{T-t}} \left\{ \ln \left(\frac{S_t}{K} \right) + (r + \frac{1}{2} \sigma^2)(T-t) \right\}$$

$$d_2 = d_1 - \sigma \sqrt{T-t}$$

Put option

$$P = e^{-r(T-t)} K N(-d_2) - S N(-d_1)$$

Forward pricing

$$0 = F(S_0, 0) = e^{-rT} E^Q(S_T - K | S_0 = S_0) \Rightarrow K = S_0 e^{rT}$$

Option on a forward

$$\left\{ \begin{array}{l} \pi(S_t) = e^{r(T-t)} \left[e^{-r(T-t)} E^Q_{\max[S_T - K e^{-r(T-t)}, 0]} \right] \\ \text{call} \xrightarrow{T-t} \text{forward} \xrightarrow{T_1} \text{European option with } K' = K e^{-r(T-t)} \end{array} \right.$$

Greeks

$$\begin{array}{lll} \Delta & \frac{\partial C}{\partial S} & \Delta^C \in [0, 1] \\ \Theta & \frac{\partial C}{\partial t} & \Theta^C > 0 \\ \nu & \frac{\partial C}{\partial \sigma} & \nu^C > 0 \\ P & \frac{\partial C}{\partial r} & P^C > 0 \\ \Gamma & \frac{\partial^2 C}{\partial S^2} & \Gamma^C = \frac{\partial \Delta^C}{\partial S_t} \quad \Gamma^P = \frac{\partial \Delta^P}{\partial S_t} \end{array}$$

t : time to maturity

Delta-hedging

For 1 call option, short $\Delta = N(d_1)$ shares of stock

In continuous time, Δ -hedging is riskless

In discrete time, gain from big movements if long call-option
 Δ ↑ as S_t ↑

Θ offsets T rent

$$\text{if } r=0 \quad \Theta = -\frac{1}{2} \sigma^2 S^2 \frac{\partial^2 C}{\partial S^2}$$

Put-Call Parity

$$C + K e^{-r(T-t)} = P + S$$

Multi-dimensional Black-Scholes

$$\begin{cases} dF = \mu_F F dt + \sigma_F F dW_t & "P" \\ \frac{\partial F}{\partial t} + \sum_{i=1}^n r S_i F_{S_i} + \frac{1}{2} \text{tr} \{ \sigma^2 F_{SS} S \sigma \} - r F = 0 \\ F(T, S_T) = \Phi(S_T) \\ \Rightarrow \left\{ \begin{array}{l} F(t, S_t) = e^{-r(T-t)} E^Q [\Phi(S_T) | S_{t+} = S_t] \\ dS_i = r S_i dt + \sigma S_i dW_i \end{array} \right. & "Q" \end{cases}$$

$$S_0 = \begin{bmatrix} S_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & S_n \end{bmatrix}$$

Admissible

$$\int_0^t h_s(u) dS(u) \geq -\alpha \quad \text{for all } (t, T]$$

First Fundamental theorem of asset pricing

no arbitrage \Leftrightarrow exists an equivalent martingale measure

Second Fundamental Theorem

Q is unique \Leftrightarrow market is complete, all claims replicated

Girsanov Theorem

$$dW^Q = \Phi_t dt + dW^Q \quad (\text{drift can be modified, not volatility})$$

$$\Rightarrow dS = S \bar{\mu}^Q dt + \sigma S (dW^Q + \bar{\rho}_t dt) \quad \text{Exists}$$

$\bar{\rho}_t = \lambda$ in the multi BS model

Incomplete market and non-tradeable assets

$$dF = \alpha_F F dt + \sigma_F F dW_t \quad \left\{ \frac{\alpha_F - r}{\sigma_F} = \frac{d_F - r}{\sigma_F} = \lambda(t) \right.$$

$$dA = \alpha_A A dt + \sigma_A A dW_t$$

$$\frac{\partial F}{\partial t} + [\mu(X_t) - \lambda(X_t) \sigma(X_t)] \frac{\partial^2 F}{\partial X_t^2} = r F(X_t)$$

$$F(T, X_T) = e^{-r(T-t)} E^Q [\Phi(X_T)]$$

$$dX_t = \{ \mu(X_t) - \lambda(X_t) \sigma(X_t) \} dt + \sigma(X_t) dW_t$$

$$\text{if traded} \quad \lambda = \frac{\mu - r}{\sigma}$$

Dividend

$$\text{discrete} \quad F_d(t, S_t) = F(t, (1-\delta)^n S_t)$$

$$\text{continuous} \quad F_d(t, S_t) = F(t, e^{-\delta(T-t)} S_t)$$

Exchange option

$$\begin{cases} dS_1 = S_1 \mu dt + \sigma_1 S_1 dW_1 \\ dS_2 = \mu S_2 dt + \sigma_2 S_2 dW_2 \\ \langle dW_1, dW_2 \rangle = P dt \end{cases}$$

B-S

$$F_t + r S_1 F_1 + r S_2 F_2 + \frac{1}{2} \sigma_1^2 S_1^2 F_{11} + \frac{1}{2} \sigma_2^2 S_2^2 F_{22} + P S_1 S_2 F_{12} - r F = 0$$

$$F = S_2 G(t, z) \quad (z = \frac{S_1}{S_2})$$

$$\Rightarrow G(t, z) + \frac{1}{2} \sigma_2^2 G_{zz}(t, z) (\sigma_1^2 + \sigma_2^2 - 2P\sigma_1\sigma_2) = 0$$

$$\{ G(t, z) = \max \{ z - 1, 0 \}$$

$$F = S_1 N(d_1) - S_2 N(d_2)$$

$$d_1 = \frac{1}{\sigma_1} \{ \ln \frac{S_1}{S_2} + \frac{1}{2} \sigma_1^2 \}$$

$$d_2 = d_1 - \sigma_1^2 \quad \sigma_1^2 = \sqrt{(\sigma_1^2 + \sigma_2^2 - 2P\sigma_1\sigma_2)(T-t)}$$

Exchange rate

$$dX_t = X_t (r^d - r^f) dt + \sigma_X X_t dW_t$$

F-forward

$$K = X_0 e^{(r^d - r^f)T}$$

Call option

$$F(t, X_t) = X_t e^{-rt} N(d_1) - e^{-r(T-t)} K N(d_2)$$

$$d_1 = \frac{1}{\sigma_X \sqrt{T-t}} \{ \ln \frac{X}{K} + (r^d - r^f + \frac{1}{2} \sigma_X^2) (T-t) \}$$

$$d_2 = d_1 - \sigma_X \sqrt{T-t} \quad (A e^{-r(T-t)} \text{ dist.})$$

$S_0 \in [S^L, S^H]$ first hits S^h , 1 dollar

$$\pi(t, x) = E^Q [e^{-rT^h} \mathbb{1}\{T^h < T^L\} | F_t]$$

$$r S_t \frac{\partial F}{\partial S_t} + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 F}{\partial S_t^2} - r F = 0$$

$$F(S^h) = 1$$

$$F(S^L) = 0$$

$$\text{guess } F(S_t) = C_1 S_t^{\alpha_1} + C_2 S_t^{\alpha_2}$$

$$\alpha_1, \alpha_2: r\alpha + \frac{1}{2} \sigma^2 \alpha (\alpha - 1) - r = 0$$

$$C_1 = \frac{(S^h)^{-\alpha_1}}{1 - (\frac{S^h}{S^L})^{\alpha_2 - \alpha_1}}$$

$$C_2 = - \frac{(\frac{S^h}{S^L})^{-\alpha_1} \cdot (S^L)^{-\alpha_2}}{1 - (\frac{S^h}{S^L})^{\alpha_2 - \alpha_1}}$$

$$F = \frac{(\frac{S_t}{S^L})^{\alpha_1} - (\frac{S^h}{S^L})^{-\alpha_1} \cdot (\frac{S_t}{S^L})^{\alpha_2}}{1 - (\frac{S^h}{S^L})^{\alpha_2 - \alpha_1}}$$

① Look-back Put Option

历史最低价的 pay-off, 例回算 $\max \{ \}$

$$X_t = \int_0^t g(s) dW_s \quad \text{martingale}$$

$$E[X(t) | F_u] = \int_0^u g(s) dW_s + E[\int_u^t g(s) dW(s) | F_u] = X(u) + 0$$

③ moment-generating function of brownian motion

$$E[e^{aW(t)}] = e^{\frac{1}{2} a^2 t} \sim N(0, \mathbb{E})$$

$$Z(t) = e^{aW(t)} \quad dZ(t) = aZ(t) dW + \frac{1}{2} a^2 Z(t) dt \quad Z(0) = 1$$

$$\int_0^t W_s dW_s = \frac{1}{2} W(t)^2 - \frac{1}{2} t$$

$$Z(t) = W(t)$$