

| | General BM | Geo BM |
|-------------------------|------------------|---|
| $\text{Var}(X_T X_t)$ | $\sigma^2(T-t)$ | $X_t^2 e^{2\mu(T-t)} (e^{\sigma^2(T-t)} - 1)$ |
| $E(X_T X_t)$ | $X_t + \mu(T-t)$ | $X_t e^{\mu(T-t)}$ |

$$E[X^2] = \text{Var}[X] + (E[X])^2$$

$$E[X^4] \neq \text{Var}[X]^2 \quad (\text{maybe})$$

VaR pros/cons:

- $P(\Pi_{t+\Delta t} - \Pi_t \leq -V_{1-\alpha}) = \alpha$
- + Simple, interpretable,
- + compare portfolios, set/monitor risk
- + determine capital bank must hold
- what when exceeded (ES)
- worse possible loss
- exponential cost w/ portfolio size (correlations)
- not additive
- diff results from diff methods
- no sensitivity to individual underlyings (Greeks)

Merton Model Pros/Cons:

- + Simple; treat equity/debt as function of options; can use BS pricing
- only default at T, single bond for debt too simple, debt given in untraded bank loans, total asset value usually not directly observable (stock not full equity, bank loans)

VaR data analysis:

- count exceeded
- $Y \sim \text{Bin}(n, p=\alpha)$, test $P(Y \geq y | \text{model})$

- default at T
- non-stochastic r
- unobservable asset value

VCV:

- $\sigma^2 = \sigma_A^2 + \sigma_B^2 + 2\rho_{AB}\sigma_A\sigma_B$, $\sigma_A^{(n)} = \text{Value}_A \times \frac{\text{Volatility}_A^{(n)}}{\sqrt{n}} \times \sqrt{n}$
- $\text{VaR}_{1-\alpha} = \sigma N^{-1}(1-\alpha)$
- assume linear combination, risk drives geo brownian, $\mu=0$

HisSim:

- $\text{VaR}_{1-\alpha} = 100\alpha$ percentile of losses, ES = avg. of times percentile exceeded

MC:

- determine risk driver, assume stochastic dynamics, determine portfolio val as function of given many sim

Probability of default, yield & spread:

- $B_0 = Xe^{-rt}$, $B_t = Xe^{-yt} = (1-\hat{p})e^{-st} B_0$, $1-\hat{p} = e^{-st}$, $s = r-y$
- Risk neutral \hat{p} for pricing, objective p for risk management

Merton Model

- $V_t = E_t + D_t$, $dV = \mu V dt + \sigma V dz$ $\xrightarrow{(\mu-r)} \hat{p}(V_T < X) = N(-d_2)$
- $E_T = \begin{cases} 0 & \text{if } V_T \leq X \\ V_T - X & \text{d/w} \end{cases} = \max(0, V_T - X) = C_X^T(V_T)$ $[\ln V_T - \ln V_0 \sim N((r - \frac{\sigma^2}{2})T, \sigma^2 T)]$
- $D_T = \begin{cases} V_T & \text{if } V_T \leq X \\ X & \text{d/w} \end{cases} = \min(V_T, X) = X - \max(X - V_T, 0) = X - P_X^T(V_T)$

+ Simple for VaR/ES

(as it's a bond)

$$\begin{pmatrix} X = D_t e^{(r+s)(T-t)} \\ D_t = X e^{-y(T-t)} \end{pmatrix}$$

Hazard Rates: $- \int_0^t h(u) du$

$$S(t) = e^{-\int_0^t h(u) du}$$

$$P[\text{die in } [t_2, t_3] | S(t_1)] = \frac{S(t_2) - S(t_3)}{S(t_1)} \quad \left(\begin{aligned} \xi(t) &= h(t) S(t) \\ P(A|B) &= \frac{P(A, B)}{P(B)} \end{aligned} \right)$$

CDS:

$$FiL = \sum SM \Delta B(0, t_i) S(t_i)$$

$$DL = \int_0^T (1-R) M B(0, u) \xi(u) du$$

$$AP = \sum SM \Delta \int_{t_{i-1}}^{t_i} \frac{u-t_{i-1}}{t_i-t_{i-1}} B(0, u) h(u) du$$

$$FiL + AP = DL$$

Comp. Martingale: $E[X_{t+\Delta} | X_t] = X_t$

Definitions:

Complete economy (attainable derivative)

Self financing portfolio: $dV_t = \phi_t dS_t + \psi_t dB_t$

Martingale Prob World: R all martingales

Fundamental theory:

arbitrage free $\Leftrightarrow \exists$ empw

complete \Leftrightarrow empw is unique for every numerical

$$\begin{aligned} &\text{Pricing Euro options:} \\ &B_t = B_0 e^{rt}, R_t = \dots \\ &K(H) = \dots \\ &\Rightarrow dS = \dots \\ &S_t = e^{-r(T-t)} \tilde{E}_t[X_T] \end{aligned}$$

Martingale pricing

$$V_t = B_t \tilde{E}_t[B_T^{-1} X_T]$$

Changing prob world:

$$dz = dz^* - K(t) dt$$

Stochastic interest rates:

$$P(t, T) = e^{-\int_t^T r(u) du} \text{ if constant } r$$

FRA:

$$r = \frac{1}{T_2 - T_1} \ln K, \quad K = \frac{P(t, T_2)}{P(t, T_1)}$$

FRNP:

$$S_t = B_t \tilde{E}_t[B_T^{-1} X_T], \quad B_T = P(t, T)$$

$$S_t = P(t, T) \tilde{E}_t[X_T], \text{ and } \tilde{E}_t[X_T] = F_t \text{ in this world}$$

bond options:

Black's Model

$$S_t = P(t, T) \tilde{E}_t[\max\{P(T, T_1) - K, 0\}] =$$

$$= P(t, T) [\tilde{E}_t[P(T, T_1)] N(d_1) - K N(d_2)]$$

$$S_t = P(t, T) [F_t N(d_1) - K N(d_2)]$$

Feynman-Kac:

$$\frac{\partial F}{\partial t} + a(x, t) \frac{\partial F}{\partial x} + \frac{1}{2} b(x, t)^2 \frac{\partial^2 F}{\partial x^2} = r(x, t) F$$

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

$$\Leftrightarrow F(x, t) = E[e^{-\int_t^T r(x, u) du} \phi(X_T) | X_t = x]$$

$$dX = a(x, t) dt + b(x, t) dz$$

long $1/T_1$ bonds, long $1/T_2$ bonds, long 1 FRA

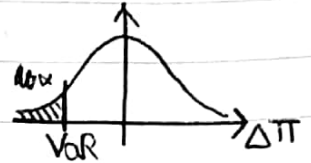
1: Market Risk Measurement

Introduction to Market Risk

- risk associated w/ uncertainty about future value/earnings due to changes in asset prices / market rates

• Value at Risk:

$$(1-\alpha) \text{ VaR: } P(\Pi_{t+\Delta t} - \Pi_t \leq -V_\alpha) = \alpha$$



Variance-Covariance Method:

1. Determine key risk drivers:

- ie what causes changes in portfolio value, eg stock prices

2. Distribution assumptions:

- (ie follows geo Brownian motion) → • risk factors follow multivariate normal distribution
(⇒ log-normally distributed changes in risk factors)

3. Sensitivity of asset value to risk factors:

- assume asset value linear combination of underlying risk drivers

4. Distribution parameters:

- assume $\mu = 0$; ie no drift (ok as focus on short time periods)
- estimate $\Sigma = \frac{1}{n-1} \sum (X_i - \bar{X})^2$ [X_i = daily returns of ith risk driver]
(or $\sigma^2 = \sigma_A^2 + \sigma_B^2 + 2\rho_{AB}\sigma_A\sigma_B$)

5. Distribution of changes in portfolio value

$$\text{VaR}_\alpha = \langle \text{Value} \rangle \times \langle n\text{-day stdev} \rangle \times N^{-1}\left(\frac{\alpha}{100}\right)$$

Historical Simulation & Monte Carlo

• VCV only works for linear products

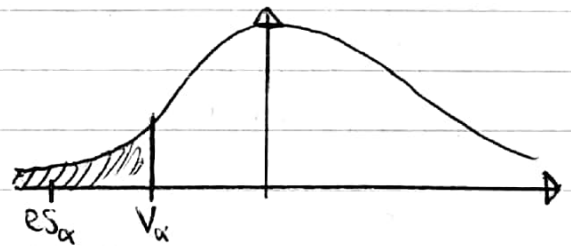
• Historical Simulation:

- determine risk drivers; get series of historic values
- convert to percentage changes ($V_i = \ln h_i / h_{i-1}$)
- form empirical distribution of portfolio P/L
- VaR is percentile of this distribution

- need lots of data to ensure we have outliers;
250+ days needed
- old data may not be representative of today
- + simple & intuitive, non-parametric (no distribution assumptions)
- Monte Carlo:
 - make assumptions on stochastic dynamics of risk drivers (including dependencies)
 - run simulation, determining distribution of portfolio value
 - time consuming to run; many simulations
 - + works well for non-linear & ~~linear~~ products

Expected Shortfall

$$-es_{\alpha}(X) = E[X | X \leq -V_{\alpha}(X)]$$



- expected loss if VaR is exceeded
- Using VCV: conditional expectations on normal distribution
- Using HisSim: mean of all times VaR exceeded

2: PCredit Risk Measurement

Introduction to Credit Risk

• risk from a party defaulting on financial obligations

• Expected loss:

$$EL = PD \times EAD \times LGD = \text{probability of default} \times \text{exposure at default} \times \text{loss given default}$$

(1-LGD = Recovery Rate) eg amount of loan still not paid back % of EAD

• Corporate bonds:

• have risk; the company may go bankrupt

• Probability of default:

• can be inferred from bond prices

$$B_G = Xe^{-rT}, \quad B_C = (1 - \hat{p})B_G \Rightarrow \hat{p} = \frac{B_G - B_C}{B_G}$$

effective rate of return on B_C (zero-coupon) risky corporate bond (zero-coupon) risk-free gov bond

• Yields & Spread:

$$B_C = Xe^{-yT} \Rightarrow \hat{p} = 1 - e^{-ST} \quad S = y - r$$

(\hat{p} is risk-neutral probability; use for pricing but not for management) risk

Counterparty Credit Risk

• Risk that counterparty defaults prior to expiration of contract

• primarily applies to OTC contracts; exchange would guarantee payments

• Incorporating into derivative pricing:

assume independent

[ξ^* = value of derivative w/ no chance of default
 \hat{p} = probability of default

$$\Rightarrow \xi = (1 - \hat{p})\xi^* = \xi^* e^{-ST}$$

ie ξ is ξ^* discounted at the spread, $y - r$
 (if $\xi^* = Xe^{-rT}$, $\xi = Xe^{-(y-r)T} = Xe^{-yT}$)

Structural Risk Credit Risk Modelling

• model firm value as stochastic process; default if falls below threshold value

• Debt & Equity in a firm:

• on bankruptcy, debt holders take priority

• equity holder benefit from increases in firm assets value

• Merton model for corporate liability valuation:

$$\begin{aligned} \text{Total value of firm assets} &= \text{Equity} + \text{Debt} \\ V(t) &= E(t) + D(t) \end{aligned}$$

default is $V_t < X \rightarrow$ • assume all debt in form of single zero-coupon bond w/ face value X & maturity time T

$$E_t = \begin{cases} 0 & \text{if } V_t < X \\ V_t - X & \text{if } V_t \geq X \end{cases} \quad \text{ie equity can be a call option, strike price } X \text{ (underlying } V_t)$$

• assume V_t follows geometric brownian motion

$$dV_t = \mu_V V_t dt + \sigma_V V_t dz$$

\Rightarrow can use Black-Scholes pricing for E

• Probability of default:

• use risk neutral world; $dV = rVdt + \sigma_V Vdz$

$$[\ln V_T - \ln V_0 \sim N(\frac{r - \sigma^2}{2} T, \sigma^2 T)] \quad \hat{p} = \hat{P}(V_T < X) = 1 - N(d_2) = N(-d_2)$$

• in real world,

$$p = P(V_T < X) = 1 - N(d_2 + \frac{\mu_V - r}{\sigma_V} \sqrt{T-t})$$

• Value of corporate debt:

$$D_T = \begin{cases} X & \text{if } V_T > X \\ V_T & \text{if } V_T \leq X \end{cases} \quad \text{ie debt can be put option, strike } X, \text{ underlying } V_t \text{ value } \min(V_T, X)$$

\Rightarrow can value by Black-Scholes, or

$$V_t = E_t + D_t \quad \min(V_T, X) = X + \max(X - V_T, 0) = X - P_X$$

\Rightarrow can value by Black-Scholes

• Or,

$$\begin{aligned} D(t) &= e^{-r(T-t)} X N(d_2) \\ &+ V_t (1 - N(d_1)) \end{aligned}$$

$$\begin{aligned} V(t) &= E(t) + D(t) = E(t) + X - P_X(t) \\ V(t) &= E(t) + D(t) \\ D(t) &= V(t) - E(t) \end{aligned}$$

(also note $E_T = V_T - D_T$)

$$E_T = V_T - X + P_{X,T} \Rightarrow E_T - P_{X,T} = V_T - X \Rightarrow C - P = V_T - X e^{-rT}$$

• Application of Merton model:

- firm's total asset value not directly observable
- stock processes only give part of information on equity
- debt often in bank loans & not traded
- need to estimate asset values & volatility
- estimate σ_V from observable equity value process, E_t , using Ito's lemma

$$\sigma_E = \frac{\partial E}{\partial V} \frac{V_t}{E_t} \sigma_V \quad (\text{as } V_t = E_t + D_t)$$

Hazard Rate Credit Models

- Treat probability (or rate) of default as exogenous variable, which may depend on firm credit rating & other variables

• Intensity models:

- model default process as point process; first point is default (inhomogeneous) \rightarrow • Poisson process, intensity (hazard rate) $h(t)$
- prob default in $(t, t+\delta t) = h(t) \delta t$

$$h(t) = \frac{f(t)}{1 - F(t)} \quad \begin{matrix} \leftarrow \text{pdfs} \\ \leftarrow \text{cdf} \end{matrix}$$

$$1 - F(t) = S(t) = \exp\left[-\int_0^t h(u) du\right]$$

↑
probability of 'survival' until time t

$$[f(t) = h(t) S(t)]$$

density instantaneous hazard change of survival to t

• Pricing credit default swaps:

- transfer of risk between parties; ~~buyer~~ ^{seller} offers to pay buyer an amount lost in case of default of some entity
- buyer pays a regular fee until default or maturity

• Fixed Leg

$$F_{FL} = E[\text{Fixed Leg}] = \sum_{i=1}^n S \Delta M B(0, t_i) S(t_i)$$

expected value of all buyer payments

$$= S M \Delta \sum_{i=1}^n B(0, t_i) e^{-\int_0^{t_i} h(u) du}$$

(n payments of SM , $\frac{1}{\Delta}$ times a year
($B(t_i, t_j)$ is value at t_i of risk free bond, maturity t_j)

• Default Leg: considering possibility of default, expected value of payment given this

$$DL = E[\text{Default Leg}] = \int_0^T (1-R) M B(0, u) g(u) du$$

$$= (1-R) M \int_0^T B(0, u) h(u) e^{-\int_0^u h(w) dw} du$$

(R = recovery rate), T = maturity)

• Accrual payment:

• seller deducts from default leg payment to account for fees from last scheduled payment until default time

$$AP = E[\text{Accrual payment}] = \sum_{i=1}^n s \Delta M \int_{t_{i-1}}^{t_i} \frac{u - t_{i-1}}{t_i - t_{i-1}} B(0, u) g(u) du$$

$$= \sum_{i=1}^n s \Delta M \int_{t_{i-1}}^{t_i} \frac{u - t_{i-1}}{t_i - t_{i-1}} B(0, u) h(u) e^{-\int_0^u h(w) dw} du$$

$$F_i L + AP = DL \Rightarrow \text{solve for } s$$

3: Asset Pricing Techniques

Fundamental Theorem of Asset Pricing

- Asset price process S_t
- Derivative price process X_t , payoff $X_T(S_T)$ at maturity
- Additional tradeable asset with process B_t (numeraire asset)
- ϕ_t of underlying, ψ_t of numeraire
 $\Rightarrow V_t = \phi_t S_t + \psi_t B_t$, want V_t to replicate X_t
- Martingales: Fair Game processes:
 - Process X_n is a fair game if $E[X_{n+1} | X_n] = X_n$ (replace n w/ t for continuous time)
 - $dS = \mu S dt + \sigma S dz$
 $\Rightarrow E[S_{t+\Delta t} | S_t] = S_t e^{\mu \Delta t}$
 $\Rightarrow S$ is fair game if $\mu = 0$; $dX = \sigma X dz$
 (also must be bounded to be fair game)
- Numeraire, relative pricing process & self-financing:
 - Numeraire:
 - tradeable asset w/ strictly +ve price process
 - can express all other asset price processes in terms of numeraire
 - $R_t = \frac{X_t}{B_t}$, relative price process
 (B_t is like measurement unit, currency if not specified)
- Self-financing:
 - if funds enter/exit our replicating portfolio, arbitrage opportunities exist
 - any change in ϕ_t must be funded by changes in ψ_t (& vice versa)
 - $\Rightarrow dV_t = \phi_t dS_t + \psi_t dB_t$
- Complete economies:
 - if self-financing strategy that replicates derivative payoff X_T exists, derivative is attainable
 - if all derivatives in an economy are attainable, economy is complete

in Black-Scholes
framework,
 $N=R=1$

- $N+1$ tradeable assets (1 risk free)
- R sources of randomness (eg independent Brownian motions)
- $N \leq R \Rightarrow$ Arbitrage free $\Rightarrow N=R \Rightarrow$ Arbitrage free & Complete
- $N \geq R \Rightarrow$ Complete

• Martingale Probability Worlds:

• Equivalent Martingale Prob World if all relative price processes of all tradeable assets are martingale processes

• Fundamental theorem of Asset pricing:

• Market arbitrage free iff exists equivalent martingale probability world measure

• Arbitrage free Market complete iff this empw is unique for every choice of numeraire

• Tells us, given choice of numeraire, can find a probability world where relative price processes are martingales, & self-financing trading strategies cannot outperform the market

by theory, exist unique Martingale probability world

Martingale Pricing Technique

- Consider complete & arbitrage free market, and numeraire with price process B_t
- Self financing trading strategy has process V_t , and $V_T = X_T$ for derivative w/ payoff X_T at maturity

$$\tilde{E} \left[\frac{V_T}{B_T} \mid V_t, B_t \right] = \tilde{E}_t \left[\frac{V_T}{B_T} \right] = \frac{V_t}{B_t}$$

expectation
in Martingale
world

$$\Rightarrow \tilde{E}_t \left[\frac{X_T}{B_T} \right] = \frac{V_t}{B_t} \Rightarrow V_t = B_t \tilde{E}_t [B_T^{-1} X_T] \quad (\text{by no arbitrage, } V_t \text{ is Value of derivative})$$

- Constructing martingale measure & process:

need to find distribution of underlying asset price under a probability measure leading to its relative price process being a martingale

- Changing probability worlds:

- Z_t is Brownian motion in probability world P
- $k(t)$ is a (possibly stochastic) process where $\int_0^t k(u)^2 du < \infty$
- k defines change to new probability world P^* ,

$$dZ = dZ^* - k(t)dt$$
 ie $Z^*(t) = Z(t) + \int_0^t k(u)du$

\Rightarrow changing probability world changes drift of process, but not volatility

\Rightarrow can think of drift as same as probability measure

Example - Black-Scholes Framework

• Underlying price process:
$$dS = \mu S dt + \sigma S dz$$

• Numeraire:

$$dB = rB dt \quad (\Rightarrow B_t = B_0 e^{rt})$$

• Relative price process:
$$R_t = S_t / B_t = S(t) e^{-rt}$$

• Find martingale probability world:

• By Ito's lemma, $dR = (\mu - r)R dt + \sigma R dz$
 $\Rightarrow k(t) = \frac{\mu - r}{\sigma}$

S_0 in world P^* , $dR = \sigma R dz^*$ ($dz = dz^* - k(t)dt$)
(0 drift, as required)

and

$$dS = rS dt + \sigma S dz^* \quad (\text{drift rate now } r)$$

providing us with probability distribution for S_T

• Pricing result:

$$S_t = B_t \tilde{E}_t [B_T^{-1} X_T]$$

$$= B_t \tilde{E}_t [(B_t e^{r(T-t)})^{-1} X_T]$$

$$S_t = e^{-r(T-t)} \tilde{E}_t [X_T] \quad [\text{risk neutral pricing result}]$$

Stochastic Interest Rate Environment

- constant r reasonable assumption for pricing equity derivatives, as uncertainty in r dominated by, eg, volatility in underlying asset / risk drivers
- not good assumption for products with greater dependence on cost of money, eg interest rate products
- Discount bonds:

- pays 1 unit of currency at maturity at time T
- follows price process $P(t, T)$ [$P(t, T) = e^{-r(T-t)}$ if r constant]

• Forward pricing:

- underlying S_t , payoff $X_T = S_T - F$

- find F through no arbitrage:

replicating portfolio \rightarrow $\pi_T = 1$ long S , F short discount bonds maturity T

$$\pi_T = S_T - F$$

$$\pi_t = S_t - F P(t, T) = 0 \quad (\text{so } \pi_t = 0, \text{ agreed forward at time } t)$$

$$\Rightarrow F = S_t / P(t, T)$$

• Forward rate agreements

- Party A pays Party B principle at time $T_1 > t$
- B pays A principle + pre-agreed interest amount at $T_2 > T_1$
- Assume A receive 1 unit at T_2 ; what should A pay at T_1 ?

- π = short R T_1 bonds, long 1 T_2 bond

$$\pi_t = R P(t, T_2) - R P(t, T_1) = 0$$

$$\pi_t = R P(t, T_2) - R P(t, T_1) = 0 \Rightarrow R = \frac{P(t, T_2)}{P(t, T_1)}$$

(A pay R at T_1 , receive 1 at T_2)

\Rightarrow agreed interest rate =

$$R e^{r(T_2 - T_1)} = 1 \Rightarrow r = \frac{1}{T_2 - T_1} \ln R$$

Example - Forward Risk Neutral Pricing

- Derivative product payoff depends on future interest rate products
 \Rightarrow cannot ignore uncertainty of future interest rates
- Many underlying assets ($P(t, T)$ for many T 's) we can use as numeraires (don't need to use one based on spot interest rate)
- Forward risk-neutral process:

$$S_t = B_t \tilde{E}_t[B_T^{-1} X_T]; \text{ hard to evaluate expectation if } B_T \text{ stochastic}$$

$$\Rightarrow \text{use } B_t = P(t, T), B_T = 1$$

$$S_t = P(t, T) \tilde{E}_t[X_T]$$

$$\text{also, as seen } F_t = S_t / P(t, T) = S_t / P(t, T) = \tilde{E}_t(X_T)$$

\Rightarrow in this world can assume expected payoff = forward value

• Bond options:

- option to buy bond at future time for agreed price
- value due to uncertain future interest rates
- Euro call option on T_2 bond, strike price K ; maturity $T < T_2$
 $X_T = \max\{P(T, T_2) - K, 0\}$

$$S_t = B_t \tilde{E}_t[S_T / B_T]$$

invalid if t close to T_2 ; close to T_2 value approaches 1

$$S_t = P(t, T) \tilde{E}_t[\max\{P(T, T_2) - K, 0\}]$$

Assume $P(t, T)$ log normally distributed

$$\Rightarrow S_t = P(t, T) (\tilde{E}[P(T, T_2)] N(d_1) - K N(d_2))$$

in formula sheet (?)
 \downarrow
 $E[\max(V-x, 0)]$
 \parallel
 $E[V] N(d_1) - x N(d_2)$

$$S_t = P(t, T) [F_t N(d_1) - K N(d_2)]$$

Black's Model

$$(d_1 = \frac{\ln(F_t/K) + \sigma^2(T-t)/2}{\sigma\sqrt{T-t}}, d_2 = d_1 - \sigma\sqrt{T-t})$$

Relationship between PDEs & SDEs

• Feynman-Kac result: (relate BSM PDE to SDE)

BSM PDE \rightarrow PDE $\frac{\partial G}{\partial t} + a(s,t) \frac{\partial G}{\partial s} + \frac{1}{2} b(s,t)^2 \frac{\partial^2 G}{\partial s^2} - r(s,t) G = 0$

payoff at maturity \rightarrow boundary conditions
 $G(s,T) = \psi(s,T)$

risk neutral pricing formula \rightarrow has solution
 $E[e^{-\int_t^T r(s,u) du} \psi(s,T) | S_t = x]$, where $ds = a(t,s)dt + b(t,s)dz$

\Rightarrow can use ^{certain} SDEs to solve ^{certain} PDEs & vice versa
 (eg MC simulation to find SDE expectations, then use to solve PDEs)

• Formulation of result:

$$dS = a(S,t)dt + b(S,t)dz = a dt + b dz, \quad S_0 = S_0, \text{ known } S_0$$

$$\Rightarrow dF = \left[\frac{\partial F}{\partial s} a + \frac{\partial F}{\partial t} + \frac{1}{2} \frac{\partial^2 F}{\partial s^2} b^2 \right] dt + \frac{\partial F}{\partial z} b dz, \text{ for } F(S,t)$$

$$\text{If } \frac{\partial F}{\partial s} a + \frac{\partial F}{\partial t} + \frac{1}{2} \frac{\partial^2 F}{\partial s^2} b^2 = 0 \Rightarrow dF = \frac{\partial F}{\partial z} b dz$$

$$\Rightarrow F(S_T, T) - F(S_t, t) = \int_t^T \frac{\partial F}{\partial z} b dz_u$$

$$\Rightarrow E[F(S_T, T) | S_t = S_0] - F(S_0, t) = E\left[\int_t^T \frac{\partial F}{\partial z} b dz_u | S_t = S_0\right]$$

$$\Rightarrow E[F(S_T, T) | S_t = S_0] = F(S_0, t) \quad [\text{rhs 0 by properties of Brownian motion}]$$

• Extension for BSM PDE:

$$rF = \frac{\partial F}{\partial s} rS + \frac{\partial F}{\partial t} + \frac{1}{2} \frac{\partial^2 F}{\partial s^2} \sigma^2 S^2, \quad F(S_T, T) = \psi(S_T)$$

(eg $\psi(S_T) = \max(S_T - K, 0)$ for a call)

• Let $G(S_t, t) = e^{r(T-t)} F(S_t, t)$

• Sub into BSM PDE

$$\Rightarrow \frac{\partial G}{\partial s} rS + \frac{\partial G}{\partial t} + \frac{1}{2} \frac{\partial^2 G}{\partial s^2} \sigma^2 S^2 = 0$$

\Rightarrow can use Feynman-Kac result,

$$G(S_0, t) = E[G(S_T, T) | S_t = S_0]$$

$$G(S_0, t) = E[\psi(S, T) | S_t = S_0] \quad (\text{where } dS = rSdt + \sigma Sdz)$$

Substitute F back in,

$$F(S_0, t) = e^{-r(T-t)} E[F(S_T, T) | S_t = S_0]$$

$$F_t = e^{-r(T-t)} \hat{E}[F_T | S_t] \leftarrow \text{risk neutral pricing formula}$$