

### UTILITY THEORY

<b>Utility function</b> $U(w)$	<b>Non-satiation:</b> $U'(w) > 0$	<b>Risk averse:</b> $U''(w) < 0$
	Risk neutral: $U''(w) = 0$	<b>Risk seeking:</b> $U''(w) > 0$
	<b>Absolute risk aversion</b>	<b>Relative risk aversion</b>
<b>Formula</b>	$A(w) = -U''(w)/U'(w)$	$R(w) = -wU''(w)/U'(w)$
<b>Increasing</b>	$A'(w) > 0$	$R'(w) > 0$
<b>Constant</b>	$A'(w) = 0$	$R'(w) = 0$
<b>Decreasing</b>	$A'(w) < 0$	$R'(w) < 0$
<b>Maximum premium</b> $P$	policy holder for random loss $X$ with initial wealth $a$	$E[U(a - X)] = U(a - P)$
<b>Minimum premium</b> $Q$	insurer for random loss $Y$ with initial wealth $a$	$E[U(a + Q - Y)] = U(a)$

### RISK MEASURES

	<b>Continuous variable</b>	<b>Discrete variable</b>
<b>Variance</b>	$\int_{-\infty}^{\infty} (\mu - x)^2 f(x) dx$	$\sum_i (\mu - x_i)^2 p_i$ for all $i$
<b>Semi-Variance</b>	$\int_{-\infty}^{\mu} (\mu - x)^2 f(x) dx$	$\sum_i (\mu - x_i)^2 p_i$ for $i: x_i < \mu$
	If the distribution is symmetric, Semi-variance = $\frac{1}{2} \times$ Variance	
<b>Value-at-Risk (VaR) at <math>p</math></b>	VaR $_p = t$ where $t$ is the 100 $p$ -th percentile, i.e. $P(X < t) = p$ . For Normal distribution: VaR $_p = \mu + Z_p\sigma$	
<b>Shortfall probability at <math>L</math></b>	$P(X < L) = \int_{-\infty}^L f(x) dx$	$P(X < L) = \sum_i p_i$ for $i: x_i < L$
<b>Expected Shortfall at <math>L</math></b>	$E[\max\{0, L - x\}] = \int_{-\infty}^L (L - x)f(x) dx$ $E[\max\{0, L - x\}] = \sum_i (L - x_i)p_i$ for $i: x_i < L$	

### ASSET VALUATIONS

#### Notations

$$E_i = E(R_i) \quad V_i = \sigma_i^2 = \text{Var}(R_i) \quad C_{ij} = \text{Cov}(R_i, R_j) \quad \rho_{ij} = \text{Corr}(R_i, R_j) = \frac{C_{ij}}{\sigma_i \sigma_j}$$

**Portfolio  $P$  of  $N$  Assets**

$$E_P = \sum_{i=1}^N x_i E_i \quad V_P = \sum_{i=1}^N x_i^2 V_i + 2 \sum_{i=1}^N \sum_{j < i} x_i x_j C_{ij} \quad V_p = \sum_i \sum_j x_i x_j C_{ij}$$

#### Minimum Variance Portfolio

Two Assets with  $\rho_{12} \neq 1$ :  $x_1 = \frac{V_2 - C_{12}}{V_1 + V_2 - 2C_{12}}$       If  $\rho_{12} = \pm 1$ :  $x_1 = \mp \frac{\sigma_2}{\sigma_1 \mp \sigma_2}$

- With  $n$ -risky assets Lagrangian:  $F(\mathbf{x}, \lambda) = \sum_{ij} x_i C_{ij} x_j - \lambda \left( \sum_i x_i - 1 \right)$  gives equations

$$2 \sum_j C_{ij} x_j - \lambda = 0 \quad \sum_i x_i = 1$$

<b>Efficient Portfolio</b>	Lagrangian given $E_P$ : $F(\mathbf{x}, \lambda) = V - \lambda(E - E_P) - \mu \left( \sum_i x_i - 1 \right)$ where $V = \sum_{ij} x_i C_{ij} x_j$
	$2 \sum_j C_{ij} x_j - \lambda E_i - \mu = 0$
	$\sum_i E_i x_i = E_P \quad \sum_i x_i = 1$ The solution $\mathbf{x}$ is linear in $E_P$
<b>Portfolio Diversification</b>	Given $x_i = \frac{1}{n}$ : $V_P = \frac{\bar{V}}{n} + \frac{n-1}{n} \bar{C}$ where $\bar{V} = \text{avg}_i[V_i]$ , $\bar{C} = \text{avg}_{i \neq j}[C_{ij}]$
<b>CAPM</b>	Risk premium = $E_P - r_f$ Market price of risk = $\phi_P = \frac{E_P - r_f}{\sigma_P}$ $\phi_M = \max_P \phi_P$
	For $P$ on the capital market line $E_P - r = \frac{E_M - r}{\sigma_M} \cdot \sigma_P$
	For any security $i$ : $E_i - r = \beta_i \cdot (E_M - r)$ where $\beta_i = \frac{\text{Cov}(R_i, R_M)}{V_M}$
<b>Single-index model</b>	$R_i = \alpha_i + \beta_i R_M + \varepsilon_i$ $R_M$ and $\varepsilon_i$ are uncorrelated.
	$\varepsilon_i$ and $\varepsilon_j$ are independent. $E_i = \alpha_i + \beta_i \cdot E_M$
	$V_i = \beta_i^2 \cdot V_M + V_{\varepsilon_i}$ $C_{ij} = \beta_i \cdot \beta_j \cdot V_M$
<b>Multi-factor model</b>	$R_i = \alpha_i + \beta_{i1} I_1 + \dots + \beta_{in} I_n + \varepsilon_i$
	$E_i = \alpha_i + \sum_j \beta_{ij} E[I_j]$
	$V_i = \sum_j \beta_{ij}^2 \text{Var}[I_j] + 2 \sum_{j < k} \beta_{ij} \beta_{ik} \text{Cov}(I_j, I_k) + V_{\varepsilon_i}$
	$C_{ij} = 2 \sum_{i < j} \beta_{ik} \beta_{jk} \text{Var}[I_k]$

**STOCHASTIC CALCULUS**

<b>Martingale process <math>X_t</math></b>	Filtration $\{\mathcal{F}_t\}_{t \geq 0}$	$E[X_s   \mathcal{F}_t] = X_t$ if $t \leq s$	$E[ X_t ] < \infty$
	Supermartingale: $E[X_s   \mathcal{F}_t] \leq X_t$	Submartingale: $E[X_s   \mathcal{F}_t] \geq X_t$	
<b>Wiener process <math>W_t</math></b>	$W_0 = 0$	$W_t - W_s \sim N(0, t - s)$	$\text{Cov}[W_s, W_t] = \min\{s, t\}$
	Disjoint increments are independent: i.e. $W_{t_2} - W_{t_1}$ and $W_{t_3} - W_{t_2}$ are independent if $t_1 < t_2 < t_3$		
<b>Itô integral</b>	$\int_0^T \sigma(W_t, t) dW_t = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} \sigma(W_{t_i}, t_i) (W_{t_{i+1}} - W_{t_i})$		
	if $\sigma(\cdot, t) \in C^2$ and $E \left[ \int_0^T \sigma^2(W_t, dt) dt \right] < \infty$		
<b>- Itô isometry</b>	$E \left[ \int_0^T \sigma(W_t, dt) dW_t \right] = 0$		
	$\text{Var} \left[ \int_0^T \sigma(W_t, dt) dW_t \right] = \int_0^T E [\sigma^2(W_t, dt)] dt$		
<b>- Martingale</b>	$E \left[ \int_0^T \sigma(W_t, t) dW_t \middle  \mathcal{F}_S \right] = \int_0^S \sigma(W_t, t) dW_t$ if $T > S$		
<b>- Linearity</b>	$\int_0^T (\sigma(W_t, t) + \nu(W_t, t)) dW_t = \int_0^T \sigma(W_t, t) dW_t + \int_0^T \nu(W_t, t) dW_t$		
<b>- Deterministic integrand</b>	$E \left[ \int_0^T \sigma(t) dW_t \right] = 0$	$\text{Var} \left[ \int_0^T \sigma(t) dW_t \right] = \int_0^T \sigma^2(t) dt$	
<b>Itô Process</b>	$X_T = X_0 + \int_0^T \mu(X_t, t) dt + \int_0^T \sigma(X_t, t) dW_t \quad dX_t = \mu(X_t, t) dt + \sigma(X_t, t) dW_t$		

**ITÔ'S LEMMA & SDE**

**Itô's lemma**

Assuming  $f(x, t) \in C_x^2 \cap C_t^1$   $dX_t = \mu_t dt + \sigma_t dW_t$

Using  $(dW_t)^2 = dt, dW_t dt = 0, dt dW_t = 0, (dt)^2 = 0:$

$$df(X_t, t) = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dX_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} (dX_t)^2$$

$$= \left( \frac{\partial f}{\partial t} + \mu_t \frac{\partial f}{\partial x} + \frac{1}{2} \sigma_t^2 \frac{\partial^2 f}{\partial x^2} \right) dt + \sigma_t \frac{\partial f}{\partial x} dW_t$$

**Arithmetic Brownian motion**

$dX_t = \mu dt + \sigma dW_t$

$X_T = X_0 + \mu T + \sigma W_T \sim N(X_0 + \mu T, \sigma^2 T)$

**Geometric Brownian motion**

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

Apply Itô's lemma with  $f(x, t) = e^x, dX_t = \left( \mu - \frac{\sigma^2}{2} \right) dt + \sigma dW_t$

$S_T = S_0 \exp \left[ \left( \mu - \frac{\sigma^2}{2} \right) T + \sigma W_T \right] \sim LN \left( \ln X_0 + \left( \mu - \frac{\sigma^2}{2} \right) T, \sigma^2 T \right)$

**Ornstein-Uhlenbeck process**

$dX_t = -\kappa X_t dt + \sigma dW_t$

Apply Itô's lemma with  $f(x, t) = x e^{\kappa t}$

$X_T = X_0 e^{-\kappa T} + \sigma \int_0^T e^{-\kappa(T-t)} dW_t \sim N \left( X_0 e^{-\kappa T}, \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa T}) \right)$

**Mean-reverting process**

$dX_t = \kappa(\theta - X_t) dt + \sigma dW_t$

Apply Itô's lemma with  $f(x, t) = x e^{\kappa t}$

$X_T = X_0 e^{-\kappa T} + \theta (1 - e^{-\kappa T}) + \sigma \int_0^T e^{-\kappa(T-t)} dW_t$

$\sim N \left( X_0 e^{-\kappa T} + \theta (1 - e^{-\kappa T}), \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa T}) \right)$

**Feller (CIR) process**

$dX_t = \kappa(\theta - X_t) dt + \sigma \sqrt{X_t} dW_t$

**INTEREST RATE MODELS**

**Interest rates**

$R(t, T)$  = effective interest rate,  $r(t, T)$  = continuous rate,

$P(t, T)$  = discount bond price at  $t$  with expiry  $T$

$P(t, T) = \frac{1}{(1 + R(t, T))^{T-t}}$

$= \exp[-(T - t)r(t, T)]$

Short rate:  $r_t = r(t, t + \delta) \approx R(t, t + \delta)$

**Forward rates**

Discrete:  $F(0; t, T) = \left( \frac{P(0, t)}{P(0, T)} \right)^{\frac{1}{T-t}} - 1$  Continuous:  $f(0; t, T) = r(0, t) + \frac{(r(0, T) - r(0, t))T}{T - t}$

**Instantaneous forward rates**

$\lim_{t \rightarrow T^-} f(0; t, T) = -\frac{\partial}{\partial T} \ln P(0, T)$   $P(t, T) = \exp \left[ -\int_t^T f(s, u) du \right]$

**Short-rate model**

$dr_t = \mu(t, r_t) dt + \sigma(t, r_t) d\hat{W}_t$

under Martingale measure  $\mathbf{Q}$

If  $P(t, T, r_t) = g(t, r_t)$ :  $g$  must satisfy:

$g(T, r) = 1$  and  $\frac{\partial g}{\partial t} + \frac{\partial g}{\partial r} \mu(t, r_t) + \frac{1}{2} \frac{\partial^2 g}{\partial r^2} \sigma^2(t, r_t) - r_t g(t, r_t) = 0$

**Vasicek's model**

$dr_t = \kappa(\theta - r_t) dt + \sigma d\hat{W}_t$

$\theta$  : mean level of short rate,  $\kappa$ : speed of reversion

$g(t, r_t) = P(t, T; r_t)$ :  $g$  must satisfy  $g(T, r) = 1$  and  $\frac{\partial g}{\partial t} + \frac{\partial g}{\partial r} \kappa(\theta - r_t) + \frac{1}{2} \frac{\partial^2 g}{\partial r^2} \sigma^2 - r_t g(t, r_t) = 0$

$P(t, T) = E \left[ \exp \left( \int_t^T r_s ds \right) \right] = \exp[H(T - t) - G(T - t)r_t]$  where:

$G(\tau) = \frac{1}{\kappa} (1 - e^{-\kappa \tau})$

$H(\tau) = \left( \theta - \frac{\sigma^2}{\kappa^2} \right) [G(\tau) - \tau] - \frac{\sigma^2}{4\kappa} G^2(\tau)$

**CIR model**

$dr_t = \kappa(\theta - r_t) dt + \sigma \sqrt{r_t} d\hat{W}_t$

$\frac{\partial g}{\partial t} + \frac{\partial g}{\partial r} \kappa(\theta - r_t) + \frac{1}{2} \frac{\partial^2 g}{\partial r^2} \sigma^2 r_t - r_t g(t, r_t) = 0$

$$P(t, T) = \exp [H(T-t) - G(T-t)r_t] \quad \text{where } \gamma = \sqrt{\kappa^2 + 2\sigma^2} \text{ and}$$

$$G(\tau) = \frac{2(e^{\gamma\tau} - 1)}{(\gamma + \kappa)(e^{\gamma\tau} - 1) + 2\gamma} \quad H(\tau) = \frac{2\kappa\theta}{\sigma^2} \ln \left( \frac{2\gamma \exp \left[ \frac{1}{2}(\gamma + \kappa)\tau \right]}{(\gamma + \kappa)(e^{\gamma\tau} - 1) + 2\gamma} \right)$$

**Hull-White model**  $dr_t = \kappa(t)(\theta(t) - r_t) dt + \sigma d\hat{W}_t$  Same as Vasicek except  $G(T-t)$ ,  $H(T-t)$  replaced with:

$$G(t, T) = \int_t^T \exp \left[ - \int_t^s \kappa(u) du \right] ds \quad H(t, T) = - \int_t^T \left( \kappa(s)\theta(s)G(s, T) - \frac{1}{2}\sigma^2 G^2(s, T) \right) ds$$

**RISK MODELS**

**Credit risk** Expected Credit Loss (ECL) = Exposure At Default (EAD) × Probability of Default (PD) × Loss Given Default (LGD)

Recovery rate = 100% – LGDProvision = PV[ECL]

**Firm valuation**  $V_t$ : firm’s value,  $D_t$ : debt,  $E_t$ : equity  $V_t = E_t + D_t$

**Merton’s model** If  $D_t = D$   $E_T = \max\{V_T - D, 0\}$   $E_t = V_t\Phi(d_1) - De^{-r(T-t)}\Phi(d_2)$

Implicit equation for  $\sigma_V$ :  $\sigma_V = \frac{\sigma_E E_t}{E_t + De^{-r(T-t)}\Phi(d_2)}$   $d_1 = \frac{\ln \frac{V_t}{D} + (r \pm \sigma_V^2/2)(T-t)}{\sigma_V \sqrt{T-t}}$

$1 - \Phi(d_2) = \Phi(-d_2)$   $d_2 = d_1 - \sigma_V \sqrt{T-t}$

Risk-neutral PD =  $\Phi(-d_2)$  For real world probabilities, replace  $\mu_V$  with  $r$  when calculating  $d_1$  and  $d_2$

**Poisson model, no recovery**  $L$ : Loan’s value if no default  $\lambda$ : Rate of default (hazard rate)  $E[\text{loan’s value}] = Le^{-\lambda T}$

**Model with recovery**  $q(t, T)$ : PD at  $t$  expiring at  $T$   $\delta$  = recovery rate  $P(t, T)$  = Loan’s value at  $t$  if no default

$$E[\text{loan’s value}] = P(t, T)[q(t, T)\delta + (1 - q(t, T))] \quad q(t, T) = \frac{1}{1 - \delta} \left( 1 - \frac{E[\text{loan’s value}]}{P(t, T)} \right)$$

**Jarrow-Lando-Turnbull model**  $Q(t, T) = k \times k$  credit-rating transition matrix with default as the  $k$ -th state  $\Lambda$  = hazard rate matrix

Diagonalisable:  $\Lambda = \Sigma D \Sigma^{-1} D = \text{Diag}[d_j]$ ,  $\Sigma = [\sigma_{ij}]$ ,  $\Sigma^{-1} = [\hat{\sigma}_{ij}]$

$$Q(t, T) = \Sigma e^{D(T-t)} \Sigma^{-1} \quad q_{ik}(t, T) = \sum_{j=1}^{k-1} \sigma_{ij} \hat{\sigma}_{jk} \left( e^{d_j(T-t)} - 1 \right)$$

If  $\Lambda = \Lambda(t) = \Sigma D U(t) \Sigma^{-1}$ :  $q_{ik}(t, T) | U(t) = \sum_{j=1}^{k-1} \sigma_{ij} \hat{\sigma}_{jk} \left( E \left[ \exp \left[ d_j \int_t^T U(s) ds \right] \right] - 1 \right)$

**LIABILITY VALUATIONS**

**General loss development model**  $C_{ij} = r_j \cdot s_i \cdot x_{i+j} + e_{ij}$   $C_{ij}$ : incremental claim in  $AY_i$ - $DY_j$

$r_j$ : development factor (df) for  $DY_j$

$s_i$ : exposure parameter for  $AY_i$

$x_{i+j}$ : parameter for  $CY_{i+j}$   $e_{ij}$ : error

**Development factors** Arithmetic avg:  $f_j = \frac{1}{n} \sum_{i=1}^n f_{ij}$

Weighted avg:  $f_j = \frac{\sum_i f_{ij} \times L_{i,j-1}}{\sum_i L_{i,j-1}}$   $f_{ij} = \frac{L_{i,j}}{L_{i,j-1}}$  where  $L_{i,j}$  is cumulative

<b>Chain-ladder (CL)</b>	$C_{ij} = r_j \cdot s_i + e_i$	Estimated ultimate claims $_i = L_{ij} \times f_{ult}$ where $f_{ult} = f_j \times f_{j+1} \times \dots$
<b>Inflation adjusted CL</b>	$C_{ij} = r_j \cdot s_i \cdot x_{i+j} + e_{ij}$	To adjust for inflation, first convert cumulative figures to incremental figures
<b>Average cost per claim (ACpC)</b>	(1) Calculate past ACpC	(2) Develop average ACpC      (3) Develop claim count
	Estimated ultimate claims $_i = \text{Ultimate ACpC}_i \times \text{Ultimate claim count}_i$	
<b>Bornhuetter-Ferguson method</b>	Estimated ultimate loss = Earned premium $\times$ Loss Ratio; Estimated reserve = Estimated ultimate loss $\times (1 - 1/f_{ult})$	

**RUIN THEORY**

<b>Aggregate claim</b>	$S(t) = \sum_{i=1}^{N(t)} X_i$	$N(t)$ : number of claims, $X_i$ : amount for $i$ -th claim, $S(t)$ : aggregate claim
<b>Surplus process</b>	$U(t) = U + ct - S(t)$	$U$ : initial surplus $c$ : premium income rate
<b>Ruin probability</b>	$\Psi(U) = P[U(t) < 0 \text{ for some } t : 0 < t < \infty]$ , $\Psi(U, \tau) = P[U(\tau) < 0 \text{ for some } t : 0 < \tau \leq t]$ $\Psi(U), \Psi(U, \tau)$ are decreasing in $U$ , increasing in $\tau$ , $\Psi(U, \tau) < \Psi(U)$ $\lim_{t \rightarrow \infty} \Psi(U, t) = \Psi(U)$	
<b>Poisson process</b>	$N(t)$ : Poisson process with rate $\lambda$ $P[N(t) = k] = \exp[-\lambda t] \frac{(\lambda t)^k}{k!}$ Inter-event time: $P(T_k > t) = \exp[-\lambda t]$	
<b>Lundberg's inequality</b>	$\Psi(U) \leq \exp[-RU]$	$R$ : adjustment coefficient      For large $U$ , $\Psi(U) \approx \exp[-RU]$
<b>Compound Poisson model</b>	$E[S(t)] = \lambda t E[X_i]$ $\text{Var}[S(t)] = \lambda t E[X_i^2]$ To make $E[U(t)] > U$ : $c = (1 + \theta)\lambda E[X_i]$ $R$ is the unique positive root of: $\lambda(M_X(r) - 1) - cr = 0$	$M_S(r) = \exp[\lambda t(M_X(r) - 1)]$ $\theta$ : premium loading factor ( $\theta > 0$ ) If $X_i \sim \text{Exp}(\alpha)$ : $R = \alpha - \frac{\lambda}{c}$
	Bounds for adjustment coefficient: $R < \frac{2(c - \lambda E[X_i])}{\lambda E[X_i^2]}$	If $X_i \leq M$ : $R > \frac{1}{M} \ln \frac{c}{\lambda E[X_i]}$
<b>General aggregate model</b>	$c > E[S_i]$ $R$ is unique positive value satisfying: $E[\exp[R(S_i - c)]] = 1$	$S_i$ : aggregate claim in year $i$ $\gamma > 0$ : $\lim_{r \rightarrow \gamma^-} E[\exp[r(S_i - c)]] = \infty$
<b>Ruin probability dependence</b>	$\Psi(U, \tau)$ decreases for larger $\theta$ If $X_i \sim \text{Exp}(1)$	For compound Poisson models: $\Psi(U, \tau)$ increases for larger $\lambda$ $\Psi(U) = \frac{1}{1 + \theta} \exp\left(-\frac{\theta U}{1 + \theta}\right)$
<b>Proportional reinsurance</b>	$c = [(1 + \theta) - (1 + \xi)(1 - \alpha)]\lambda E[X_i]$ $\xi$ : reinsurer premium loading factor Reinsurer constraint: $\alpha > \frac{\xi - \theta}{1 + \xi}$ If $\theta = \xi$ , then $\alpha > 0$ . i.e. any retention level is possible.	$\alpha$ : retention level Primary insurer constraint: $\alpha > 1 - \frac{\theta}{\xi}$ $\xi > \theta$
<b>Excess of loss reinsurance</b>	$c = (1 + \xi)\lambda E[Y_i] - (1 + \theta)\lambda E[Z_i]$	

Reinsurer loss:	$Z_i = \max\{0, X_i - M\}$
Primary insurer loss:	$Y_i = \min\{X, M\}$
$R$ satisfies:	$\lambda + cR = \lambda \left( \int_0^M e^{Rx} f_X(x) dx + e^{RM}(1 - F_X(M)) \right)$
	$M = \text{retention level}$

**OPTION THEORY**

**Notations**  $S_t / B_t$  : share/bond price  $K$ : strike  $T$ : expiry  $\phi_t / \psi_t$ : number of shares/bond

$c_t / p_t$ : European call/put price

$C_t / P_t$ : American call/put price

$r$ : risk-free rate

$\sigma$ : share price volatility  $\delta$ : continuous dividend yield  $I$ : fixed dividend at  $t = 0$

**Self-financing strategy**  $(\phi_t, \psi_t)$  satisfying:  $V(t) = \phi_t S_t + \psi_t B_t$   $dV(t) = \phi_t dS_t + \psi_t dB_t$

**Replicating strategy** Self-financing  $(\phi_t, \psi_t)$  satisfying: where  $X_T$  is the derivative payoff  
 $\phi_T S_T + \psi_T B_T = X_T$

**Option payoff** European call:  $\max(S_T - K, 0)$  European put:  $\max(K - S_T, 0)$

Option bounds	Option Type	Upper bound	Lower bound
	European call	$c_t \leq S_t$	$c_t \geq S_t - K \exp[-r(T - t)]$
	American call	$C_t \leq S_t$	$C_t \geq S_t - K \exp[-r(T - t)]$
	European put	$p_t \leq K \exp[-r(T - t)]$	$p_t \geq K \exp[-r(T - t)] - S_t$
	American put	$P_t \leq K$	$P_t \geq K - S_t$

Parameter-price relationship	Increased parameter	Call price	Put price
	Strike ( $K$ )	Decrease	Increase
	Time to expiry ( $T - t$ )	Increase	Increase
	Volatility ( $\sigma$ )	Increase	Increase
	Risk-free rate ( $r$ )	Increase	Decrease
	Dividend ( $\delta$ )	Decrease	Increase
	Share price ( $S_t$ )	Increase	Decrease

**Forward price** zero dividend:  $S_0 e^{rT}$  fixed dividend:  $(S_0 - I)e^{rT}$  continuous dividend:  $S_0 \exp[(r - \delta)T]$

**One-period replicating portfolio**

$$\Delta = \phi_0 = \frac{V_u - V_d}{S_0 e^{\delta h}(u - d)} \quad B = \psi_0 = \frac{uV_d - dV_u}{e^{rh}(u - d)} \quad V_0 = \Delta S_0 + B$$

**Risk-neutral probability**  $Q_{(\text{price increase})}$   $V_0 = e^{-rh} E_Q[V(n)]$   $Q$ : Risk-neutral Measure  
 $= q = \frac{e^{(r-\delta)h} - d}{u - d}$

**Calibrating binomial model**  $u = \exp[\sigma\sqrt{h} + \delta dt]$   $d = \exp[-\sigma\sqrt{h} + \delta dt]$   $h \rightarrow 0$ :  
 $\ln \frac{S_t}{S_0} \sim N\left(\left(r - \frac{\sigma^2}{2}\right)t, \sigma^2 t\right)$

**State-price deflator** ( $\delta = 0$ )

$$A_1 = e^{-r} \frac{q}{p} \text{ if } S_1 = S_0 u \quad A_1 = e^{-r} \frac{1-q}{1-p} \text{ if } S_1 = S_0 d \quad V_0 = E_P[A_1 V_1]$$

$$A_n = e^{-rn} \left(\frac{q}{p}\right)^{N_n} \left(\frac{1-q}{1-p}\right)^{n-N_n}$$

$N_n$ : number of up's til time  $n$

$$V_0 = E_P[A_n V_n]$$

### BLACK-SCHOLES MODEL

**Equivalent measure**

$P \sim Q$  are equivalent  $\iff P(E) > 0$  whenever  $Q(E) > 0$  for any event  $E$ .

**Cameron-Martin-Girsanov**

There exists an equivalent measure  $Q$  s.t.  $W_t + \int_0^t \gamma_s ds$  is a Wiener process, where  $\gamma_s$  is a previsible process.

**Martingale discounted price**

To make  $e^{-rt} S_t$  martingale, apply CMG with  $\gamma_t = \frac{\mu - r}{\sigma}$ ,

$$\text{where } S_t = S_0 \exp \left[ \left( \mu - \frac{\sigma^2}{2} \right) t + \sigma W_t \right]$$

**Martingale representation**

$X_t, Y_t$ : martingale. Then,  $dY_t = \phi_t dX_t$  for some  $\phi_t$

$\iff X_t$  is not martingale under any other equivalent measure.

**Black-Scholes formula** ( $\delta = 0$ )

$$c_t = S_t N(d_1) - K e^{-r(T-t)} N(d_2) \quad p_t = K e^{-r(T-t)} N(-d_2) - S_t N(-d_1)$$

$$d_1 = \frac{\ln(S_t/K) + (r + \frac{1}{2}\sigma^2)(T-t)}{\sigma\sqrt{T-t}} \quad d_2 = d_1 - \sigma\sqrt{T-t}$$

**Black-Scholes Equation** ( $\delta = 0$ )

$$\frac{\partial f}{\partial t} + rS_t \frac{\partial f}{\partial s} + \frac{\sigma^2 S_t^2}{2} \frac{\partial^2 f}{\partial s^2} = rf(t, S_t) \quad f(T, s) = (s - K)_+ \text{ for call, } (K - s)_+ \text{ for put}$$

**BSE derivation** ( $\delta = 0$ )

$V(t, S_t)$ : portfolio value with

$f = f(t, S_t)$ : option value

1 derivative and  $\frac{\partial f}{\partial s}$  shares

$$\implies dV_t = -df + \frac{\partial f}{\partial s} dS_t$$

Applying Itô's lemma:

Arbitrage-free implies:

$$dV_t = \left( -\frac{\partial f}{\partial t} - \frac{1}{2} \frac{\partial^2 f}{\partial s^2} \sigma^2 S_t^2 \right) dt$$

$$dV_t = rV_t dt$$

**Martingale approach** ( $\delta = 0$ )

$F_t$ : filtration,

Find equivalent measure  $Q$

$X$ : contingent derivative payment

s.t.  $D_t = e^{-rt} S_t$  is martingale

$$f(t, S_t) = \exp[-r(T-t)] E_Q[X | F_t]$$

Then,  $E_t = e^{-rt} f(t, S_t)$  is martingale

For some  $\phi_t$ :  $dE_t = \phi_t dD_t$

Let  $\psi_t = E_t - \phi_t D_t$

(martingale representation)

$(\phi_t, \psi_t)$  is self-financing.

Derivative value:  $f(t, S_t) = \phi_t S_t + \psi_t B_t$

**Martingale to BSE** ( $\delta = 0$ )

$$dE_t = e^{-rt} (-rf(t, S_t) dt + df)$$

$$= \frac{\partial f}{\partial s} dD_t + e^{-rt} \left[ -rf(t, S_t) + \frac{\partial f}{\partial t} + rS_t \frac{\partial f}{\partial s} + \frac{\sigma^2 S_t^2}{2} \frac{\partial^2 f}{\partial s^2} \right] dt$$

$$\text{Martingale representation: } dE_t = \phi_t dD_t \implies \phi_t = \frac{\partial f}{\partial s}$$

$$\text{Martingale} \implies -rf(t, S_t) + \frac{\partial f}{\partial t} + rS_t \frac{\partial f}{\partial s} + \frac{\sigma^2 S_t^2}{2} \frac{\partial^2 f}{\partial s^2} = 0$$

**State-price deflator approach** ( $\delta = 0$ )

$$dS_t \stackrel{P}{=} \mu S_t dt + \sigma S_t dZ_t$$

$P \sim Q$ : By CMG,  $d\tilde{Z}_t = dZ_t + \gamma_t dt$

$$\stackrel{Q}{=} rS_t dt + \sigma S_t d\tilde{Z}_t$$

with  $\gamma_t = \frac{\mu - r}{\sigma}$

$$\eta_t = \exp \left[ -\gamma Z_t - \frac{1}{2} \gamma^2 t \right]$$

State-price deflator:  $A_t = e^{-rt} \eta_t$

$$f(t, S_t) = e^{-r(T-t)} \mathbb{E}_Q[X|F_t] = e^{-r(T-t)} \mathbb{E}_P \left[ \frac{\eta_T}{\eta_t} X \middle| F_t \right] = \frac{\mathbb{E}_P[A_T X | F_t]}{A_t}$$

**BSE derivation** ( $\delta > 0$ )

$\tilde{S}_t$ : value of investment starting with  $S_0$   $d\tilde{S}_t = (\mu + \delta)\tilde{S}_t dt + \sigma dZ_t$

$$\tilde{S}_t = \tilde{S}_0 \exp \left[ \left( \mu + \delta - \frac{\sigma^2}{2} \right) t + \sigma Z_t \right] \quad \text{Change in derivation:}$$

$$dV_t = -df + \frac{\partial f}{\partial s} (dS_t + \delta S_t dt)$$

$$\frac{\partial f}{\partial t} + (r - \delta) S_t \frac{\partial f}{\partial s} + \frac{\sigma^2 S_t^2}{2} \frac{\partial^2 f}{\partial s^2} = r f(t, S_t)$$

**Black-Scholes formula** ( $\delta > 0$ )

$$c_t = S_t e^{-\delta(T-t)} N(d_1) - K e^{-r(T-t)} N(d_2)$$

$$p_t = K e^{-r(T-t)} N(-d_2) - S_t e^{-\delta(T-t)} N(-d_1)$$

$$d_1 = \frac{\ln(S_t/K) + (r - \delta + \frac{1}{2}\sigma^2)(T-t)}{\sigma\sqrt{T-t}} \quad d_2 = d_1 - \sigma\sqrt{T-t}$$

**Option Greeks**

$$\Delta = \frac{\partial f}{\partial s}, \quad \Gamma = \frac{\partial^2 f}{\partial s^2}, \quad \Theta = \frac{\partial f}{\partial t},$$

BSE ( $\delta = 0$ ):

$$\nu = \frac{\partial f}{\partial \sigma}, \quad \rho = \frac{\partial f}{\partial r}, \quad \lambda = \frac{\partial f}{\partial \delta}$$

$$\Theta + rs\Delta + \frac{1}{2}\sigma^2 s^2 \Gamma = r f$$

$$\Delta_{\text{call}} = e^{-\delta(T-t)} N(d_1)$$

$$\Delta_{\text{put}} = -e^{-\delta(T-t)} N(-d_1)$$