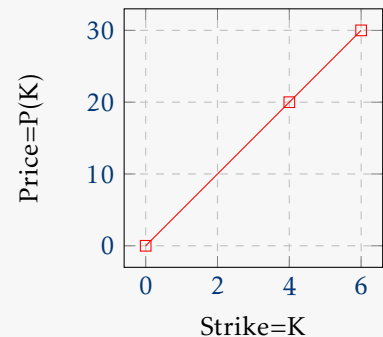


## 1 2 put options, K=20,30 trading for \$4,\$6 - find arbitrage

- $2x6-3x4=0$ , so try  $2(20-S)^+ - 3(30-S)^+ > 0$
- $\begin{cases} S \geq 30 \Rightarrow 0 \\ 20 \leq S < 30 \Rightarrow -ve \\ 0 \leq S < 20 \Rightarrow -ve \end{cases}$
- 0 or -ve, so take opposite and done
- graph (K,P(K)) not **strictly** convex  $\Rightarrow$  arbitrage



## 2 $2^{29}$ has 9 digits all different - find missing digit

- key: 9 divides  $[n - \text{Sum}(\text{digits } n)]$
- **Proof:**  $n = a_n 10^n + a_{n-1} 10^{n-1} + \dots + a_1 10^1 + a_0$
- $S(n) = a_n + a_{n-1} + \dots + a_1 + a_0$
- $n - S(n) = a_n(10^n - 1) + a_{n-1}(10^{n-1} - 1) + \dots + a_1(10^1 - 1) + a_0 - a_0$
- $10^k - 1 =$  all 9s = divisible by 9. so QED
- $\text{Sum}(\text{digits } n) = \sum_{i=1}^9 i - x = 45 - x$
- so  $9|(2^{29} - 45 - x)$
- but  $2^{29} = 2^5 \times 2^{6 \times 4} = 2^5 \times 64^4 = 2^5 \times (63 + 1)^4 = 2^5 \times (k \times 63 + 1) = k \times 63 \times 2^5 + 2^5 = 9 \times \text{something for some } k$
- so  $9|(2^{29} - 45 - x)$  and  $9|(2^{29} - 2^5)$  so  $9|(2^{29} - 45 - x) - (2^{29} - 2^5)$
- $9|(45 - x - 2^5) = (45 - x - 32) = (13 - x)$  so  $x = 4$

## 3 $\int_0^T W_t dt$ - what is distribution? is it martingale?

- $X_T = \int_0^T W_t dt \Leftrightarrow dX_t = W_t dt + 0 \cdot dW_t =$  only drift so not martingale
- $dX_t = W_t dt \Leftrightarrow X_T = \int_0^T W_t dt \Leftrightarrow X_T = [W_t t]_0^T - \int_0^T t dW_t = TW_T - \int_0^T t dW_t = T \int_0^T dW_t - \int_0^T t dW_t = \int_0^T (T-t) dW_t$
- Recall stochastic integral  $\int_0^T f(t) dW_t \sim N(0, \int_0^T |f(t)|^2 dt)$
- $X_T = \int_0^T (T-t) dW_t \sim N(0, \int_0^T |T-t|^2 dt) =$

$N(0, \frac{T^3}{3})$

## 4 Alice sends 20 ants towards Bob in straight line, Bob 50 towards Alice in straight line. Ants collide and go back. How many reach Bob? Alice? How many collisions?

- imagine ants carry a flag and pass it on. so 50 reach Alice and 20 reach Bob.
- num collisions: each of Bob's 50 flags must go through 20 collisions to reach Alice, each of Alice's 20 flags goes through 50 collisions to reach Bob:  $2 \times 50 \times 20 = 2000$  collisions

## 5 find all $p$ such that this is a correlation matrix

$$\Omega = \begin{bmatrix} 1 & 0.6 & -0.3 \\ 0.6 & 1 & p \\ -0.3 & p & 1 \end{bmatrix}$$

- $\Omega$  correlation matrix  $\Leftrightarrow \Omega$  symmetric positive definite
- **Sylvester, Cholesky, definition**
- **Sylvester** criterion: all the principal minor determinants must be  $> 0$  (principal minor: remove the same rows and same column indices form a matrix)

$$1. \text{ remove } 2,3 \rightarrow \det(1) = 1, \text{ remove } 1,3 \rightarrow \det(1) = 1, \text{ remove } 1,2 \rightarrow \det(1) = 1$$

$$2. \text{ remove } 3 \rightarrow \det \begin{bmatrix} 1 & 0.6 \\ 0.6 & 1 \end{bmatrix} = 0.64$$

$$3. \text{ remove } 2 \rightarrow \det \begin{bmatrix} 1 & -0.3 \\ -0.3 & 1 \end{bmatrix} = 0.91$$

$$4. \text{ remove } 1 \rightarrow \det \begin{bmatrix} 1 & p \\ p & 1 \end{bmatrix} = 1 - p^2$$

$$5. \text{ remove } \begin{bmatrix} 1 & 0.6 & -0.3 \\ 0.6 & 1 & p \\ -0.3 & p & 1 \end{bmatrix} \rightarrow \det \Omega = \det \begin{bmatrix} 1 & 0.6 & -0.3 & 0.6 & 0.6 \\ 0.6 & 1 & p & 0.6 & 1 \\ -0.3 & p & 1 & -0.3 & p \end{bmatrix} \rightarrow + \text{down-up} \rightarrow 1 - 0.18p - 0.18p - 0.09 - p^2 - 0.36 = 0.55 - 0.36p - p^2$$

- $\begin{cases} 1 - p^2 > 0 \\ 0.55 - 0.36p - p^2 > 0 \end{cases}$

## 6 how many draws $N$ of $U[0,1]$ such that $P(0.7 < U[0,1] < 0.72) \geq 0.95$ for one $U[0,1]$

- $P(\text{none of the } N \text{ rv in } [0.7, 0.72]) = 0.98^N \rightarrow P(\text{at least one in } [0.7, 0.72]) = 1 - 0.98^N$  and solve  $1 - 0.98^N \geq 0.95$

## 7 prove pdf of $N(0,1)$ is: $\int \text{pdf} = 1$

- $\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} \exp(-\frac{x^2}{2}) dx$  substitute  $x^2/2 = t \rightarrow dx = \sqrt{2} dt \rightarrow$  need prove  $I := \int_{-\infty}^{+\infty} \exp(-t^2) dt = \sqrt{\pi}$
- $I^2 = \int_{-\infty}^{+\infty} \exp(-x^2) dx \int_{-\infty}^{+\infty} \exp(-y^2) dy = \int_{-\infty}^{+\infty} \exp(-x^2 + y^2) dx dy$
- $x = r \cos \theta, y = r \sin \theta, dx dy = r dr d\theta$

$$I^2 = \int_0^{+\infty} \int_0^{+2\pi} \exp(-r^2) r dr d\theta = 2\pi \int_0^{+\infty} \exp(-r^2) r dr = 2\pi \left[ -\frac{1}{2} \exp(-r^2) \right]_0^{+\infty} = \pi \rightarrow I = \sqrt{\pi}$$

## 8 walk 1m south, 1m east, 1m north, then back at same point - where are you on Earth?

- find latitude near south Pole such that round-the-world=1m. then any point 1m north of this works
- also any latitude near south Pole such that round-the-world= $\frac{1}{k}$  works (go round the world  $k$  times but still end up same point)

## 9 solve Ornstein-Uhlenbeck SDE - Vasicek for IR

$$\bullet dr_t = \lambda(\theta - r_t)dt + \sigma dW_t \rightarrow dr + \lambda r_t dt = \lambda \theta dt + \sigma dW_t \xrightarrow{\times e^{\lambda t}} e^{\lambda t}(dr + \lambda r_t dt) = e^{\lambda t}(\lambda \theta dt + \sigma dW_t) \rightarrow d(r_t e^{\lambda t}) = e^{\lambda t}(\lambda \theta dt + \sigma dW_t) \rightarrow [r_t e^{\lambda t}]_0^t = \int_0^t \lambda \theta e^{\lambda s} ds + \int_0^t \sigma e^{\lambda s} dW_s \rightarrow r_t e^{\lambda t} - r_0 = \theta(e^{\lambda t} - 1) + \sigma \int_0^t e^{\lambda s} dW_s \rightarrow r_t = e^{-\lambda t} r_0 + \theta(1 - e^{-\lambda t}) + \sigma \int_0^t e^{-\lambda(t-s)} dW_s \text{ and that is it.}$$

- note however  $E(\int_0^t f(s) dW_s) =$  "sum of  $N(0, \cdot)$ "=0 so  $E(r_t) = e^{-\lambda t} r_0 + \theta(1 - e^{-\lambda t})$  which  $\rightarrow \theta$  as  $r \uparrow + \infty$  "mean reverting"

## 10 Calculus

- $i^i? e^{i \frac{\pi}{2} i} = e^{-\frac{\pi}{2}}$
- $\pi^e > e^{\pi}?$  try  $\ln(\pi^e) = e \ln(\pi)$  vs  $\ln(e^{\pi}) = \pi \ln(e)$  ie  $\frac{\ln(\pi)}{\pi}$  vs  $\frac{\ln(e)}{e}$  ie check  $f(x) = \frac{\ln(x)}{x}$  ie  $f'(x) = (\frac{\ln(x)}{x})' = \frac{1}{x} \frac{1}{x} + \ln(x) \frac{-1}{x^2} = \frac{1 - \ln x}{x^2}$  ie  $f'(x) < 0$  if  $x > e$  and  $f'(x) > 0$  if  $x < e$  ie  $f(x) \uparrow$  on  $[0, e]$  and  $f \downarrow$  on  $[e, \dots + \infty]$  and  $e < \pi$  so  $f(e) > f(\pi)$
- $\frac{e^x + e^y}{2} > e^{\frac{x+y}{2}}? e^{\frac{x+y}{2}} = \sqrt{e^{x+y}} = \sqrt{e^x e^y}$  ie  $a^2 + b^2 > 2ab$  ie  $a^2 - 2ab + b^2 > 0$  which is true
- solve  $x^6 = 64$   $2^6 = 64$  and  $z^6 = 1 \Leftrightarrow z = e^{\frac{2ik\pi}{6}}, k \in 0 \dots 5$  so  $x = 2e^{\frac{2ik\pi}{6}}, k \in 0 \dots 5$
- derivative of  $x^x? x^x = e^{x \ln(x)}$  so derivative =  $g' \circ f \cdot f'$  ie  $e^{x \ln(x)} (\ln(x) + x \frac{1}{x}) = x^x (\ln(x) + 1)$
- compute  $\sqrt{2 + \sqrt{2 + \sqrt{2} \dots}}? l = \sqrt{2 + l}$  ie  $l^2 = l + 2$  ie  $(l-2)(l+1) = 0$  ie  $l = 2$  since  $l > 0$  assuming limit exists. but  $x_{n+1} = \sqrt{x_n + 2}$  is increasing (same reason) and bounded above, so  $l$  exists and  $l=2$
- find  $2 = x^{x^{x^{\dots}}}$   $2 = x^2$  so  $x = \sqrt{2}$  if it exists. prove  $\sqrt{2} = \sqrt{2}^{\sqrt{2}^{\sqrt{2}^{\dots}}}$  let  $x_0 = \sqrt{2}, x_{n+1} = x_n^{\sqrt{2}}$ . Prove sequence is increasing, bounded from above by 2. therefore has a limit. Therefore

that limit must be  $\sqrt{2}$

- which one converges?  $\sum \frac{1}{k}, \sum \frac{1}{k^2}, \sum \frac{1}{k \ln(k)}$

$$1. \sum \frac{1}{k^2} \leq \sum \frac{1}{k(k-1)} = \sum \frac{1}{k} - \frac{1}{k-1} \leq 1 - 1/n < 2$$
 so  $\uparrow$ , bounded ie converges.

$$2. \sum \frac{1}{k} > \ln(n) + \frac{1}{n} \text{ because } \int_1^n \frac{1}{x} dx = \sum_{i=1}^{n-1} \int_k^{k+1} \frac{1}{x} dx < \sum_{i=1}^{n-1} \int_k^{k+1} \frac{1}{k} dx = \sum_{i=1}^{n-1} \frac{1}{k} [x]_k^{k+1} = \sum_{i=1}^{n-1} \frac{1}{k} = \sum_{i=1}^n \frac{1}{k} - \frac{1}{n}$$

$$3. \sum \frac{1}{k \ln(k)} : \text{similarly compare } \int_1^n \frac{1}{x \ln(x)} dx \text{ and note } \int \frac{1}{x \ln(x)} = \ln(\ln(x))$$

$$\bullet \text{ compute } \int \frac{1}{1+x^2} dx? \text{ substitute } x = \tan(z) = \frac{\sin z}{\cos z} \text{ so } dx = \frac{1}{\cos^2 z} dz = \cos z \frac{1}{\cos^2 z} + \sin z \frac{-1}{\cos^2 z} (-\sin z) = 1 + \frac{\sin^2 z}{\cos^2 z} = \frac{1}{\cos^2 z} \text{ so } \int \frac{1}{1+x^2} dx = \int \frac{1}{1+\tan^2 z} \frac{1}{\cos^2 z} dz = \int \frac{1}{\cos^2 z + \sin^2 z} dz = \int dz = z + C = \arctan x + C$$

$$\bullet \text{ compute } \int x \ln(x) dx? \int x \ln(x) dx = [\frac{x^2}{2} \ln x] - \int \frac{x^2}{2} \frac{1}{x} dx$$

$$\bullet \text{ compute } \int x e^x dx? \int x e^x dx = [x e^x] - \int e^x dx$$

$$\bullet \text{ compute } \int x^n \ln x dx \int x^n \ln x dx = [\frac{x^{n+1}}{n+1} \ln x] - \int \frac{x^{n+1}}{n+1} \frac{1}{x} dx$$

$$\bullet \text{ compute } \int \ln^n x dx? \int \ln^n x dx = [x \ln^n x] - \int x n \ln^{n-1}(x) \frac{1}{x} dx = [x \ln^n x] - n \int \ln^{n-1}(x) dx \text{ so recursion } f_n(x) = [x \ln^n x] - n f_{n-1}(x)$$

$$\bullet \text{ solve } y'' - 4y' + 4y = 1 \text{ general solution } y'' - 4y' + 4y = 0 \rightarrow z^2 - 4z + 4 = 0 \rightarrow (z-2)^2 = 0 \rightarrow y = C_1 e^{2x} + C_2 x e^{2x} + \text{particular solution } y = \frac{1}{4}$$

$$\bullet \text{ solve } y' = y(1-y) \quad y' = y(1-y) \rightarrow \frac{y'}{y(1-y)} = \frac{y'}{y(1-y)} \Rightarrow \int \frac{y'}{y(1-y)} dx = \int 1 dx \quad y' = dy/dx \Rightarrow \int \frac{dy}{y(1-y)} = \int 1 dx \text{ with } \int \frac{dy}{y(1-y)} = \int (\frac{1}{y} + \frac{1}{y-1}) dy \text{ so } \ln y - \ln(1-y) = \ln(\frac{y}{1-y}) = x + C \text{ so } \frac{y}{1-y} = e^{x+C}$$

$$\bullet \text{ derive BS PDE set } \Pi = V - \frac{\partial V}{\partial S} S \text{ so } d\Pi = dV - \frac{\partial V}{\partial S} dS \text{ and } \frac{dS}{S} = \mu dt + \sigma dW \text{ and } dV = \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} dS + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} (dS)^2 \text{ and } (dS)^2 = \sigma^2 S^2 dt \text{ so } dV = (\frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2) dt + \frac{\partial V}{\partial S} dS \text{ so } d\Pi$$

$$= \left( \frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2 \right) dt \stackrel{\text{risk free growth}}{=} r \Pi dt$$

$$= r(V - \frac{\partial V}{\partial S} S) dt \text{ so } \left( \frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2 \right) = r(V - \frac{\partial V}{\partial S} S)$$

$$\text{rearrange } \frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2 + r \frac{\partial V}{\partial S} S - rV = 0$$

## 11 Linear algebra

- cov  $\Sigma_X$  & corr  $\Omega_X$  matrices are +ve sem.-def

$$\Sigma(j, k) = \text{cov}(X_j, X_k) = \text{cov}(X_k, X_j) = \Sigma_X(k, j) \text{ and } \Omega(j, k) = \text{corr}(X_j, X_k) = \text{corr}(X_k, X_j) = \Omega(k, j).$$

Also

$$(\star) \text{var}(\sum_1^n c_i X_i) = C^T \Sigma_X C \text{ with } C = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$$

and  $\text{var}(\cdot) \geq 0$  so  $\Sigma_X$  is positive, and symmetric (see above), so +ve semi-definite

- Proof of ( $\star$ ): let  $Y = \sum_1^n c_i X_i$  then  $Y - E[Y] = \sum_1^n c_i (X_i - \mu_i)$  and  $\text{Var}(Y)$

$$= E[Y - E(Y)]^2$$

$$= E[\sum_1^n c_i (X_i - \mu_i)^2]$$

$$= E[\sum_{1 \leq j, k \leq n} c_j c_k (X_j - \mu_j)(X_k - \mu_k)]$$

$$= \sum_{1 \leq j, k \leq n} c_j c_k E[(X_j - \mu_j)(X_k - \mu_k)]$$

$$= \sum_{1 \leq j, k \leq n} c_j c_k \text{cov}(X_j, X_k)$$

$$= \sum_{1 \leq j, k \leq n} c_j c_k \Sigma_X(j, k)$$

$$= C^T \Sigma_X C$$

- For correlation:  $\Sigma_X = D_{\sigma_X} \Omega_X D_{\sigma_X}$  where

$$D_{\sigma_X} = \text{Diag}(\sigma_i), \sigma_i^2 = \text{var}(X_i), i \in 1..n$$

$$\text{and } w^T \Sigma_X w = w^T D_{\sigma_X} \Omega_X D_{\sigma_X} w = v^T D_{\sigma_X}^{-1} D_{\sigma_X} \Omega_X D_{\sigma_X} D_{\sigma_X}^{-1} v = v^T \Omega_X v \geq 0$$

so correlation  $\Omega_X$  positive semi-definite.

- Find correlation  $\Omega$  for covariance matrix

$$\Sigma = \begin{bmatrix} 0.36 & 0.36 & -1.144 \\ 0.36 & 0.8 & 0.8 \\ -1.144 & 0.8 & 0.9 \end{bmatrix}$$

$$\Sigma_X = \begin{bmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_2 & 0 \\ 0 & 0 & \sigma_3 \end{bmatrix} \Omega_X = \begin{bmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_2 & 0 \\ 0 & 0 & \sigma_3 \end{bmatrix}$$

$$\Omega_X = \begin{bmatrix} \frac{1}{\sigma_1} & 0 & 0 \\ 0 & \frac{1}{\sigma_2} & 0 \\ 0 & 0 & \frac{1}{\sigma_3} \end{bmatrix} \Sigma_X = \begin{bmatrix} \frac{1}{\sigma_1} & 0 & 0 \\ 0 & \frac{1}{\sigma_2} & 0 \\ 0 & 0 & \frac{1}{\sigma_3} \end{bmatrix} \text{ with}$$

$$\sigma_1 = \sqrt{\Sigma(1,1)} = 1, \sigma_2 = \sqrt{\Sigma(2,2)} = 2, \sigma_3 = \sqrt{\Sigma(3,3)} = 3$$

$$\text{Find allowable } p \text{ for } \Omega = \begin{bmatrix} 1 & p & \dots & p \\ p & 1 & \ddots & p \\ \vdots & \ddots & \ddots & p \\ p & p & p & 1 \end{bmatrix}$$

- Find eigenvalues of  $\Omega$  then state that all e.v must be +ve to find the condition on p
- $\Omega = (1-p)I + pM$  with  $I$  = identity.  $M$ =all 1s. But ev of  $M$  are easy to find:  $Mv = \lambda v \rightarrow v_1 + v_2 + \dots + v_n = \lambda v_1 = \lambda v_2 = \dots = \lambda v_n$  so  $\begin{cases} \lambda = 0 \\ v_1 = v_2 = v_n \rightarrow v_1 + v_2 + v_n = n v_1 = \lambda v_1 \end{cases}$  so  $\lambda = 0, n$  and for  $\Omega v = (1-p)Iv + pMv = (1-p)v + p\lambda v = (1-p+p\lambda)v$  so  $\begin{cases} \lambda = 0 \rightarrow 1-p \geq 0 \\ \lambda = n \rightarrow 1-p+pn \geq 0 \end{cases}$  so  $\frac{1}{1-n} \leq p \leq 1$

- prove  $n \times n$  matrix has  $n$  eigenvalues

- $Av = \lambda v \Leftrightarrow (\lambda I - A)v = 0 \Leftrightarrow (\lambda I - A)$  singular  $\Leftrightarrow \det(\lambda I - A) = 0 \Leftrightarrow P_A(\lambda) = 0$  so  $n$  roots, so  $n$  eigenvalues but some may be the same
- An eigenvalue of multiplicity  $m$  has at least 1 eigenvector and at most  $m$  linearly independent eigenvectors.

- find  $X^2 = A$  and  $YY^T = A$  for  $A = \begin{bmatrix} 2 & -2 \\ -2 & 5 \end{bmatrix}$

- A symmetric!  $\Leftrightarrow A = O^T \Lambda O$  with  $\Lambda = \text{Diag}(\text{evals})$  and  $O$  orthogonal and made up of the vectors ie

$$A = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}, O = (v_1 v_2),$$

$$Av_1 = \lambda_1 v_1, Av_2 = \lambda_2 v_2, \|v_1\| = 1, \|v_2\| = 1$$

- $X = O^T \Lambda^{\frac{1}{2}} O, O^T O = I$  as  $O$ =orthogonal

- A symmetric +ve definite  $\Leftrightarrow$  Cholesky  $\Leftrightarrow A = U^T U$  with  $U = \begin{bmatrix} u_1 & u_2 \\ u_3 & 0 \end{bmatrix}$  so  $Y = U^T$

- find the evals:

$$\det(\lambda I - A) = \det \begin{bmatrix} \lambda - 2 & 2 \\ 2 & \lambda - 5 \end{bmatrix} = 0$$

- find the evs with  $\| \cdot \| = 1$

- find the Cholesky  $U$  by solving linear equation  $\begin{bmatrix} u_{11} & 0 \\ u_{12} & u_{22} \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} \\ 0 & u_{22} \end{bmatrix} = A$

- if  $\lambda_1 = 2, \lambda_2 = -3, v_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, v_2 = \begin{bmatrix} -1 \\ 3 \end{bmatrix}$

$$\text{find } Av, v = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

- find  $c_1, c_2: v = c_1 v_1 + c_2 v_2$  (solve equation)
- $Av = c_1 \lambda_1 v_1 + c_2 \lambda_2 v_2 = c_1 2 v_1 + c_2 (-3) v_2$

- show  $\text{trace}(AB) = \text{trace}(BA)$  for  $A, B = n \times n$

$$(\star) P_{AB}(\lambda) = P_{BA}(\lambda) = \det(\lambda I - AB) = \det(\lambda I - BA)$$

- $P_M(\lambda) = \det(\lambda I - M) = \lambda^n - \text{trace}(M) + \dots + (-1)^n \det(M)$
- so  $\text{trace}(AB) = \text{trace}(BA)$
- Proof  $\star$ :  $\det(\lambda I - AB) = \det(\lambda I B^{-1} B - B^{-1} B A B) = \det(B^{-1}) \det(\lambda I B - B A B) = \det(B^{-1}) \det(\lambda I - B A) \det(B) = \det(B^{-1}) \det(B) \det(\lambda I - B A) = \det(B^{-1} B) \det(\lambda I - B A) = \det(\lambda I - B A)$
- for singular  $B$ , use  $B - \epsilon I$  which is non-singular apart from finite  $\epsilon = \text{evalue}$  so let  $\epsilon \rightarrow 0$

- solve  $AB - BA - I = 0$  for  $A, B = n \times n$

- no solution since  $\text{trace}(AB - BA) = \text{trace}(AB) - \text{trace}(BA) = 0$

- show  $A, B$  prob. Mat  $\Rightarrow AB$  prob. Mat

- A prob. Mat := sum of rows = 1  $\Leftrightarrow A \mathbb{I} = \mathbb{I}$

$$\text{with } \mathbb{I} = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$$

$$(AB)\mathbb{I} = A(B\mathbb{I}) = A\mathbb{I} = \mathbb{I}$$

- all  $\rho: \Omega := \begin{bmatrix} 1 & 0.6 & -0.3 \\ 0.6 & 1 & \rho \\ -0.3 & \rho & 1 \end{bmatrix} = \text{corr matrix}$

- Sylvester, Cholesky, Definition

- Sylvester**: all the principal minors must be  $> 0$  (principal minors: remove same row and same column, compute determinant) - see above

- Cholesky**: the determinant of the  $2 \times 2$  matrix  $M$  in the first step of the Cholesky algorithm must be  $\geq 0$ :

$$M = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix} - \begin{bmatrix} 0.6 \\ -0.3 \end{bmatrix} \begin{bmatrix} 0.6 & -0.3 \end{bmatrix}$$

$$= \begin{bmatrix} 0.64 & \rho + 0.18 \\ \rho + 0.18 & 0.91 \end{bmatrix}$$

$$\det M = \det \begin{bmatrix} 0.64 & \rho + 0.18 \\ \rho + 0.18 & 0.91 \end{bmatrix} \geq 0$$

$$\text{Definition } x^T \Omega x \geq 0 \quad \forall x \text{ ie } \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} 1 & 0.6 & -0.3 \\ 0.6 & 1 & \rho \\ -0.3 & \rho & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \geq 0$$

$$\text{ie } x_1^2 + x_2^2 + x_3^2 + 1.2x_1x_2 - 0.6x_1x_3 + 2\rho x_1x_3 = x_1^2 + x_2^2 + x_3^2 + 1.2x_1x_2 - 0.6x_1x_3 + 2\rho x_1x_3 = x_1^2 + 1.2x_1x_2 - 0.6x_1x_3 + x_2^2 + x_3^2 + 2\rho x_1x_3 = x_1^2 + 2x_1(0.6x_2 - 0.3x_3) + x_2^2 + x_3^2 + 2\rho x_1x_3 = (x_1 + 0.6x_2 - 0.3x_3)^2 - (0.6x_2 - 0.3x_3)^2 + x_2^2 + x_3^2 + 2\rho x_1x_3 = (x_1 + 0.6x_2 - 0.3x_3)^2 + 0.64x_2^2 + 2x_2x_3(\rho + 0.18) + 0.91x_3^2 \text{ and complete last}$$

sqaure again same method

## 12 Financial instruments

- 3 put options (strike=K, Price(K)) = (40,10), (50,20), (70,30). Is there arbitrage if so build it
- graph  $K \mapsto \text{Price}(K)$  must be **strictly convex** else arbitrage
- line through (40,10) and (70,30):  $C: K \mapsto \frac{70-K}{30} 10 + \frac{K-40}{30} 30 = 10 + \frac{x-40}{70-40} (30-10) = 10 + \frac{30-10}{70-40} (x-40)$
- on that line  $P(50) = 20 > C(50) = \frac{70-50}{30} 10 + \frac{50-40}{30} 30 = 50/3$  so not convex (convex: curve must be below line) so arbitrage exists
- construct arbitrage: find portfolio in the puts so that 1.  $V(0) < 0$ , 2.  $V(T) > 0 \forall T$  with  $V(T) =$
- try  $2x_{10} + 30 - 3x_{20} = 2P_{40} + P_{50} - 3P_{70} = 2(40-K)^+ + (50-K)^+ - 3(70-K)^+$
- Price of a stock now  $P=50$ . In 3m, either  $P=47$  or  $P=52$  with prob 50-50. How much for ATM put? Assume no dividends and Interest rate = 0
- real world probability irrelevant
- standard solutions follows:
- $P(0) = p_{\text{up}} P_{\text{up}} + p_{\text{down}} P_{\text{down}} = p P_{\text{up}} + (1-p) P_{\text{down}} = pu P(0) + (1-p) d P(0)$
- $P(0) = pu P(0) + (1-p) d P(0)$
- $1 = pu + (1-p)d$  with  $u = \frac{P_{\text{up}}}{P(0)} = \frac{52}{50}, d = \frac{47}{50}$
- $1 = p(u-d) + d \Rightarrow p = \frac{1-d}{u-d} = \dots = 0.6$
- $O(0) = p O_{\text{up}} + (1-p) O_{\text{down}}$
- (here)  $O$ =ATM put, ATM means strike=Price now = 50
- $O(0) = p \cdot 0 + (1-p) \cdot 3 = 0.6 \times 0 + 0.4 \times 3 = 1.2$
- alternative solution follows:
- Set up portfolio  $\Pi = +1$  Option +  $(-1) \Delta_{\text{Option}}$  Stock with  $\Delta_{\text{Option}} = \frac{O_{\text{up}} - O_{\text{down}}}{S_{\text{up}} - S_{\text{down}}} = (\text{here}) \frac{0-3}{52-47} = 0.6$
- (here)  $\Pi = +1$  Option +  $(-1)(-0.6)$  Stock
- $\Pi(T) = \begin{cases} S(T) = 52 \Rightarrow 0 + 0.6 \times 52 = 31.2 \\ S(T) = 47 \Rightarrow 3 + 0.6 \times 47 = 31.2 \end{cases}$
- $\Pi(0) = \text{discounted } \Pi(T) \stackrel{\text{IR}=0}{=} \Pi(T) = 31.2 = O(0) + 0.6 \times S(0) = O(0) + 0.6 \times 50 \Rightarrow O(0) = 31.2 - 0.6 \times 50 = 31.2 - 30 = 1.2$
- What is risk neutral pricing?
- $V(0) = \mathbb{E}[e^{-rT} V(S(T))]$
- with  $S(T) = S(0)e^{(r-q-\frac{\sigma^2}{2})T + \sigma\sqrt{T}N(0,1)}$   $\star$

- not OK for path-dependent
- How to derive BS ? ✖
  - 12 possible ways actually ...
  - risk neutral pricing
  - BS PDE:
    - \* Payoff = boundary conditions
    - \* Transform to heat equation
  - binomial tree, with calibration:
    - \* drift = risk free rate
    - \* terminal dist = lognormal when num time steps  $\rightarrow \infty$
- Approximate formula for ATM put ? ✖
  - Put<sub>ATM</sub>  $\approx 0.4S_0\sigma\sqrt{T}$  when total variance  $= \sigma^2 T$  small
- if the price of a stock doubles, how does call option change ?
  - depends if call option ITM, ATM, OTM
  - ITM:
    - \*  $C - P = S - K \Leftrightarrow$  Put-call parity
    - \*  $C + Ke^{-rT} = S \Rightarrow C = S - Ke^{-rT}$  so  $C \times 2$
  - ATM: call option  $\rightarrow$  ITM,  $C \uparrow (\times 10)$
  - OTM: call option  $\rightarrow$  ATM,  $C \uparrow \uparrow (\times 10^n)$
- what are the possible values of Delta of an option ?
  - Call option:
    - \* long Call: [0 (OTM) ... 1 (ITM)]
  - Put option:
    - \*  $C + P = S \Leftrightarrow$  Put-Call parity
    - \*  $\Delta_C + \Delta_P = 1$
- what is Delta of +1 ATM Call ? what is Delta of +1 ATM Put ?
  - Delta of +1 ATM Call = 0.5
  - Delta of +1 ATM Put = -0.5
  - $\Delta_{Call} = N(d_1)$  with  $d_1 = \frac{\ln \frac{S}{K} + (r - d + \frac{\sigma^2}{2})}{\sigma\sqrt{T}}$  ✖
    - \*  $ATM \rightarrow K = S$ , assume  $r=q=0 \rightarrow d_1 = \frac{\sigma\sqrt{T}}{2}$
    - \*  $\Delta_{Call} = N(d_1) = N(0.5\sigma\sqrt{T})$
    - \*  $0.5\sigma\sqrt{T}$  normally small eg for  $\sigma = 0.5, T = 1 \Rightarrow 0.5 \times 0.5\sqrt{1} = 0.25$
    - \*  $N(x)$  about 0 is  $N(x) = 0.5 + \frac{x}{\sqrt{2\pi}}$  ✖
    - \* so  $\Delta_{Call} = N(d_1) \approx 0.5$  since  $d_1 = 0.5\sigma\sqrt{T}$  small
  - $\Delta_{Put} = -N(-d_1)$  ✖
    - \* so  $\Delta_{Put} = -N(-d_1) \approx -0.5$
  - $N(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{z^2}{2}} dz = 0.5 + \frac{1}{\sqrt{2\pi}} \int_0^x e^{-\frac{z^2}{2}} dz \approx 0.5 + \frac{1}{\sqrt{2\pi}} x$

- what is Put-Call parity ?
  - Call - Put = Forward
  - $(S - K)^+ - (K - S)^+ = S - K$
  - $C(0) - P(0) = [S(0)e^{rT} - K]e^{-rT}$
  - $C(0) - P(0) = S(0) - Ke^{-rT}$
  - $C(t) - P(t) = S(t) - Ke^{-r(T-t)}$
- Show that the time value of an option is highest ATM ✖
  - time value of call option := call option value - intrinsic value :=  $C(t) - (S_t - K)^+$
  - at any time  $t$ , for which value of  $S$  is this highest ?  $S = K$  ?
  - fix  $t$ , define:  $f(S) = C(S) - (S - K)^+$ 
    - \*  $S \leq K \Rightarrow f(S) = C(S)$
    - \*  $S > K \Rightarrow f(S) = C(S) - (S - K)$
    - \*  $S \leq K \Rightarrow$  call option so  $\uparrow$  with  $S$
    - \*  $S > K \Rightarrow f'(S) = \Delta_C - 1 < 0$
  - so max at  $S = K$
- What is implied volatility ? volatility smile ? volatility skew ?
  - implied volatility = the  $\sigma : BS(..., \sigma) =$  observed price. Unique  $\sigma$  because  $\sigma \uparrow \Rightarrow BS(..., \sigma) \uparrow$
  - Volatility smile  $\smile : \sigma_{ITM} > \sigma_{ATM} < \sigma_{OTM}$  [currency options]
  - Volatility skew  $\frown : \sigma_{ITM} < \sigma_{ATM} > \sigma_{OTM}$  [index options, equity options, com. options]
- What is the Gamma of an option ? Why is it better to have small Gamma ? Why is Gamma of plain vanilla options positive ?
  - Gamma  $\Gamma = \frac{\partial^2 V}{\partial S^2}$
  - small Gamma  $\Rightarrow$  Delta does not change quickly  $\Rightarrow$  easier to keep Delta-neutral
  - Delta of Call/Put  $\uparrow$  with  $\uparrow S$  ✖, hence Gamma +ve
  - Gamma  $\approx 0$  for OTM / deep ITM, Gamma max ATM ✖
- When is Call = Put ?
  - $C_0 - P_0 = (S_0 e^{rT} - K)e^{-rT} = S_0 - Ke^{-rT} = 0 \Leftrightarrow K = S_0 e^{rT}$
  - so  $C_0 = P_0 \Rightarrow K =$  forward value of asset
- What is 2-year volatility of an asset with 30 % 6m volatility ?
  - $\sigma(t) = \sigma_{1Y} \sqrt{t}$
  - $\sigma(6m) = \sigma_{1Y} \sqrt{0.5}$
  - $\sigma(2y) = \sigma_{1Y} \sqrt{2} = \sigma(6m) \frac{\sqrt{2}}{\sqrt{0.5}}$
- Value fix/float swap ?
  - PV = fix - float (say)
  - fix pv = sum of future flows
  - float = right after payment, sum of all

- flows = 1 (notional)
  - so float = pv of next payment + pv of (1) at next payment date ✖
  - example: 6m swap,  $N=10M, K=3\%$ , next payment in 1m, then 7m, 13m, 19m, with last reset saying next flt payment = 125k
    - \* fix cpn =  $Nr\tau = 10M \times 0.03 \times 0.5 = 150k$  with  $\tau =$  year fraction
    - \* fix pv =  $150k \times df(\frac{1}{12}) + 150k \times df(\frac{7}{12}) + 150k \times df(\frac{13}{12}) + (10M + 150k) \times df(\frac{19}{12})$
    - \* float pv =  $(10M + 125k) \times df(\frac{1}{12})$
    - \* example df for semi-annual LIBOR =  $df(t) = [(1 + L \times 0.5) \times (1 + L \times 0.5)]^{-t} \Leftrightarrow$  start with 1, keep on re-investing. Divide 1/(what you get) to get the df = price of zero coupon delivering 1 at T
- Price change of a 10y ZC bond if yield increases by 10bp ?
  - Duration:  $\Delta B = \frac{\partial B}{\partial y} dy = - \left( \overbrace{-\frac{1}{B} \frac{\partial B}{\partial y}}^{\text{Duration}} \right) B dy := -DBdy$  ✖
  - so  $\frac{\Delta B}{B} = -Ddy$
  - (here)
    - \* 10y ZC  $\Rightarrow D=10$
    - \*  $dy = 10bps = 10 \times 1e-4 = 1e-3$
    - \*  $\frac{\Delta B}{B} = -Ddy = -10 \times 1e-3 = -0.01$
    - \* price  $\downarrow 10\%$
- A 5y ZC bond with Duration  $D=3.5y$  has  $P=102$ . What is P if yield  $\downarrow 50bp$  ?
  - $\frac{\Delta B}{B} = -Ddy$
  - (here)
    - \*  $D = 3.5$
    - \*  $dy = -50bp = -5 \times 1e-3$
    - \*  $B = 102$
    - \*  $\Delta B = -3.5 \times 102 \times (-5 \times 1e-3) \approx +1.785$
    - \*  $B_{new} \approx B + 1.785 = 102 + 1.785 = 103.785$
- What is a forward contract ?
  - (long position) agrees to buy an asset at a specific price at specific time in future
  - (short position) agrees to sell an asset at a specific price at specific time in future
  - that specific price is called the forward price
  - at  $T=0$  the value of the forward contract is 0  $\Leftrightarrow$  obv price of the forward contract  $\neq$  forward price !! (completely unrelated quantities)
  - Forward Price =  $F = S_0 e^{(r-q)T} =$  price of the asset in the future
- Forward price of Treasury FUTURES contract vs Forward price of a Commodity FU-

- TURES contract ?
  - For Treasury (ie bond) FUTURES contract, price now includes the bond flows between now and T
  - But you won't receive these when you get the bond, so for you, price of bond must exclude them
  - ie Forward price =  $F_{bond} = \{PV_{now}(Bond) - PV_{now}(Coupons)\}e^{rT}$
  - For Com future, situation is reversed when you get the Com, you will not have incurred storage costs etc
  - ie Forward price =  $F_{com} = \{PV_{now}(Com) + PV_{now}(Storage)\}e^{rT}$
- Difference between Future and Forward ?
  - daily settlement for futures, not so for forward
  - Standardised maturities, contracts for futures, not so for forwards
  - Exchange involved for futures (reduced CPTY risk), not so for forwards (OTC)
  - Range of delivery dates for (COM) futures, not so for forwards
  - Forwards can be cash-settled, even for COMs
- 10-day VAR @ 99% of a portfolio with 5-day VAR @ 95% = USD 100M ?
  - $VAR(N, C) = \sigma_{V,1Y} P(Z \leq C) \sqrt{\frac{N}{252}} V(0)$  with
    - \*  $N =$  time horizon eg  $N=10$ -days
    - \*  $P = P$  for Normal distribution
    - \*  $C =$  confidence level = eg 95 % = 0.95
    - \*  $\sigma_{V,1Y} =$  annualised std dev of the portfolio PNL
    - \*  $V(0) =$  PV now of portfolio
  - eliminate the  $\sigma_{V,1Y} V(0)$  by division ...

### 13 C++

- declare an array
  - `T foo [3]`
  - `T bar [] = {1,2,3}`
  - `int T* baz = new int[3]`
- declare an array of pointers
  - `T* foo [3]`
  - `T* bar [] = {&a,&b,&c}`
  - `int T** baz = new int* [3]`
- const pointer to object, const pointer to const object etc
  - Remember the clockwise and backward rule
  - `T* bar [10]; // array of 10 pointers to T`

- `char const *chptr;` // pointer to a const char == pointer to a read-only char
- `char * const chptr;` // const pointer to a char == read-only pointer to a char
- `char const * const chptr;` // const pointer to a const char == read-only pointer to a read-only char

- **template functions**

```
template <typename T>
```

```
T temp_sum (T a, T b) return {a+b;}
```

```
// accepts 2 Ts, return a+b , which is also a T
```