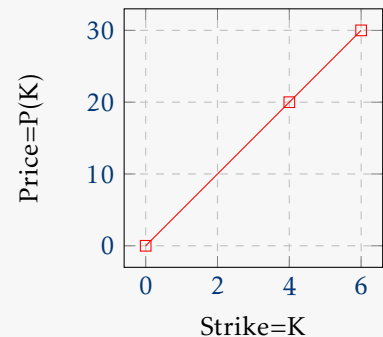


1 2 put options, K=20,30 trading for \$4,\$6 - find arbitrage

- $2x6-3x4=0$, 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:** $n = a_n 10^n + a_{n-1} 10^{n-1} + \dots + a_1 10^1 + a_0$
- $S(n) = a_n + a_{n-1} + \dots + a_1 + a_0$
- $n - S(n) = a_n(10^n - 1) + a_{n-1}(10^{n-1} - 1) + \dots + a_1(10^1 - 1) + a_0 - a_0$
- $10^k - 1 = \text{all 9s} = \text{divisible by 9. so QED}$
- $\text{Sum}(\text{digits } n) = \sum_{i=1}^9 i - x = 45 - x$
- 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?

- $X_T = \int_0^T W_t dt \Leftrightarrow dX_t = W_t dt + 0 \cdot dW_t = \text{only drift so not martingale}$
- $dX_t = W_t dt \Leftrightarrow X_T = \int_0^T W_t dt \Leftrightarrow X_T = [W_t t]_0^T - \int_0^T t dW_t = TW_T - \int_0^T t dW_t = T \int_0^T dW_t - \int_0^T t dW_t = \int_0^T (T-t) dW_t$
- 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

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

- Ω 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)

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

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

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

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

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

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

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

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

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 $E(\int_0^t f(s) dW_s) = \text{"sum of } N(0, \dots) = 0$ 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

- $i^i? e^{i \frac{\pi}{2} i} = e^{-\frac{\pi}{2}}$
- $\pi^e > e^{\pi}? \text{ try } \ln(\pi^e) = e \ln(\pi) \text{ vs } \ln(e^{\pi}) = \pi \ln(e) \text{ ie } \frac{\ln(\pi)}{\pi} \text{ vs } \frac{\ln(e)}{e} \text{ ie check } f(x) = \frac{\ln(x)}{x} \text{ ie } f'(x) = (\frac{\ln(x)}{x})' = \frac{1}{x} \frac{1}{x} + \ln(x) \frac{-1}{x^2} = \frac{1 - \ln x}{x^2} \text{ ie } f'(x) < 0 \text{ if } x > e \text{ and } f'(x) > 0 \text{ if } x < e \text{ ie } f(x) \uparrow \text{ on } [0, e] \text{ and } f \downarrow \text{ on } [e, \dots + \infty] \text{ and } e < \pi \text{ so } f(e) > f(\pi)$
- $\frac{e^x + e^y}{2} > e^{\frac{x+y}{2}}? e^{\frac{x+y}{2}} = \sqrt{e^{x+y}} = \sqrt{e^x e^y} \text{ ie } a^2 + b^2 > 2ab \text{ ie } a^2 - 2ab + b^2 > 0 \text{ which is true}$
- 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 $\sqrt{2 + \sqrt{2 + \sqrt{2} \dots}}? l = \sqrt{2 + l} \text{ ie } l^2 = l + 2 \text{ ie } (l - 2)(l + 1) = 0 \text{ ie } l = 2 \text{ since } l > 0 \text{ assuming limit exists. but } x_{n+1} = \sqrt{x_n + 2} \text{ is increasing (same reason) and bounded above, so } l \text{ exists and } l = 2$
- 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? $\sum \frac{1}{k}, \sum \frac{1}{k^2}, \sum \frac{1}{k \ln(k)}$

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

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

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

- compute $\int \frac{1}{1+x^2} dx?$ substitute $x = \tan(z) = \frac{\sin z}{\cos z}$ so $dx = \frac{1}{\cos^2 z} dz = \frac{1}{\cos^2 z} dz$ so $\int \frac{1}{1+\tan^2 z} \frac{1}{\cos^2 z} dz = \int \frac{1}{\cos^2 z} dz = \int dz = z + C = \arctan x + C$
- compute $\int x \ln(x) dx? \int x \ln(x) dx = [\frac{x^2}{2} \ln x] - \int \frac{x^2}{2} \frac{1}{x} dx$
- compute $\int x e^x dx? \int x e^x dx = [x e^x] - \int e^x dx$
- compute $\int x^n \ln x dx \int x^n \ln x dx = [\frac{x^{n+1}}{n+1} \ln x] - \int \frac{x^{n+1}}{n+1} \frac{1}{x} dx$
- compute $\int \ln^n x dx? \int \ln^n x dx = [x \ln^n x] - \int x n \ln^{n-1}(x) \frac{1}{x} dx = [x \ln^n x] - n \int \ln^{n-1}(x) dx$ so recursion $f_n(x) = [x \ln^n x] - n f_{n-1}(x)$
- solve $y'' - 4y' + 4y = 1$ general solution $y'' - 4y' + 4y = 0 \rightarrow z^2 - 4z + 4 = 0 \rightarrow (z-2)^2 = 0 \rightarrow y = C_1 e^{2x} + C_2 x e^{2x}$ + particular solution $y = \frac{1}{4}$
- solve $y' = y(1-y)$ $y' = y(1-y) \rightarrow \frac{y'}{y(1-y)} = \frac{dy}{y(1-y)}$ $1 \rightarrow \int \frac{y'}{y(1-y)} dx = \int 1 dx \xrightarrow{y'=dy/dx} \int \frac{dy}{y(1-y)} = \int 1 dx$ with $\int \frac{dy}{y(1-y)} = \int (\frac{1}{y} + \frac{1}{1-y}) dy$ so $\ln y - \ln(1-y) = \ln(\frac{y}{1-y}) = x + C$ so $\frac{y}{1-y} = e^{x+C}$
- derive BS PDE set $\Pi = V - \frac{\partial V}{\partial S} S$ so $d\Pi = dV - \frac{\partial V}{\partial S} dS$ and $\frac{dS}{S} = \mu dt + \sigma dW$ and $dV = \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} dS + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} (dS)^2$ and $(dS)^2 = \sigma^2 S^2 dt$ so $dV = (\frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2) dt + \frac{\partial V}{\partial S} dS$ 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 Σ_X & corr Ω_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 Σ_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 Ω_X positive semi-definite.

- Find correlation Ω for covariance matrix

$$\Sigma = \begin{bmatrix} 1 & 0.36 & -1.144 \\ 0.36 & 1 & 0.8 \\ -1.144 & 0.8 & 1 \end{bmatrix}$$

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

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

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

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

- Find eigenvalues of Ω 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 $A, B = n \times n$

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

$$P_M(\lambda) = \det(\lambda I - M) = \lambda^n - \text{trace}(M) + \dots + (-1)^n \det(M)$$

$$\text{so } \text{trace}(AB) = \text{trace}(BA)$$

$$\text{Proof } \star: \det(\lambda I - AB) = \det(\lambda I B^{-1} B - B^{-1} B A B) = \det(B^{-1}) \det(\lambda I B - B A B) = \det(B^{-1}) \det(\lambda I - B A) \det(B) = \det(B^{-1}) \det(B) \det(\lambda I - B A) = \det(B^{-1} B) \det(\lambda I - B A) = \det(\lambda I - B A)$$

- 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 $\Leftrightarrow A \mathbb{I} = \mathbb{I}$

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

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

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

- 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 2×2 matrix M in the first step of the Cholesky algorithm must be ≥ 0 :

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

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

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

- Definition** $x^T \Omega x \geq 0 \quad \forall x$ 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. $V(T) > 0 \quad \forall T$ with $V(T) =$

$$\text{try } 2x_{10} + 30 - 3x_{20} = 2P_{40} + P_{50} - 3P_{70} = 2(40-K)^+ + (50-K)^+ - 3(70-K)^+$$

- 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:

$$P(0) = p_{\text{up}} P_{\text{up}} + p_{\text{down}} P_{\text{down}} = p P_{\text{up}} + (1-p) P_{\text{down}} = pu P(0) + (1-p) d P(0)$$

$$P(0) = pu P(0) + (1-p) d P(0)$$

$$1 = pu + (1-p)d \text{ with } u = \frac{P_{\text{up}}}{P(0)} = \frac{52}{50}, d = \frac{47}{50}$$

$$1 = p(u-d) + d \Rightarrow p = \frac{1-d}{u-d} = \dots = 0.6$$

$$O(0) = p O_{\text{up}} + (1-p) O_{\text{down}}$$

$$\text{(here) } O = \text{ATM put, ATM means strike=Price now} = 50$$

$$O(0) = p 0 + (1-p) 3 = 0.6 \times 0 + 0.4 \times 3 = 1.2$$

- alternative solution follows:

$$\text{Set up portfolio } \Pi = +1 \text{ Option} + (-1) \Delta_{\text{Option}} \text{ Stock with } \Delta_{\text{Option}} = \frac{O_{\text{up}} - O_{\text{down}}}{S_{\text{up}} - S_{\text{down}}} = \text{(here) } \frac{0-3}{52-47} = 0.6$$

$$\text{(here) } \Pi = +1 \text{ Option} + (-1)(-0.6) \text{ Stock}$$

$$\Pi(T) = \begin{cases} S(T) = 52 \Rightarrow 0 + 0.6 \times 52 = 31.2 \\ S(T) = 47 \Rightarrow 3 + 0.6 \times 47 = 31.2 \end{cases}$$

$$\Pi(0) = \text{discounted } \Pi(T) \stackrel{\text{IR}=0}{=} \Pi(T) = 31.2 = O(0) + 0.6 \times S(0) = O(0) + 0.6 \times 50 \Rightarrow$$

$$O(0) = 31.2 - 0.6 \times 50 = 31.2 - 30 = 1.2$$

- What is risk neutral pricing?

$$V(0) = \mathbb{E} \left[e^{-rT} V(S(T)) \right]$$

$$\text{with } S(T) = S(0) e^{(r-q-\frac{\sigma^2}{2})T + \sigma \sqrt{T} N(0,1)}$$



- 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_{ATM} $\approx 0.4S_0\sigma\sqrt{T}$ 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 $\Delta_{Put} = -N(-d_1) \approx -0.5$
 - $N(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{z^2}{2}} dz = 0.5 + \frac{1}{\sqrt{2\pi}} \int_0^x e^{-\frac{z^2}{2}} dz = 0.5 + \frac{1}{\sqrt{2\pi}} \int_0^x [1 + \dots] dz \approx 0.5 + \frac{1}{\sqrt{2\pi}} x$

- 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 σ : $BS(\dots, \sigma) =$ observed price. Unique σ because $\sigma \uparrow \Rightarrow BS(\dots, \sigma) \uparrow$
 - 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 ≈ 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 τ = 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: $\Delta B = \frac{\partial B}{\partial y} dy = -\left(\overbrace{-\frac{1}{B} \frac{\partial B}{\partial y}}^{\text{Duration}}\right) B dy := -DB dy$ ✖
 - so $\frac{\Delta B}{B} = -D dy$
 - (here)
 - * 10y ZC $\Rightarrow D=10$
 - * $dy = 10bps = 10 \times 1e-4 = 1e-3$
 - * $\frac{\Delta B}{B} = -D dy = -10 \times 1e-3 = -0.01$
 - * price $\downarrow 10\%$
- A 5y ZC bond with Duration $D=3.5y$ has $P=102$. What is P if yield $\downarrow 50bp$?
 - $\frac{\Delta B}{B} = -D dy$
 - (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 ...